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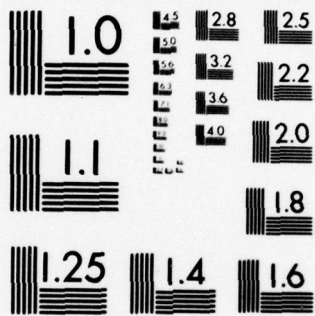
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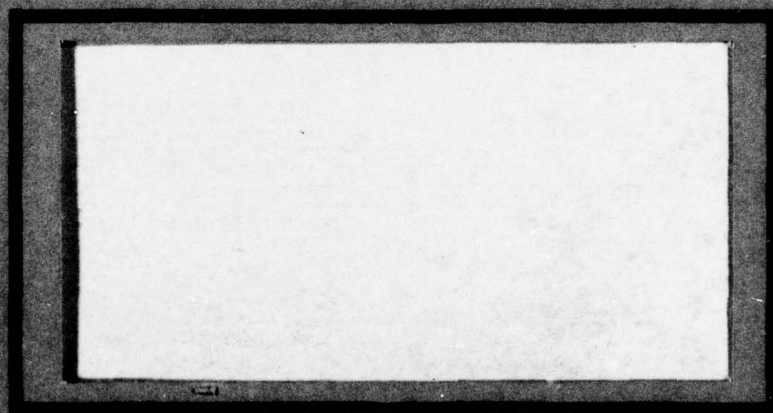
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⑥ CRITERION FOR SELECTION OF  
VARIABLES IN A REGRESSION  
ANALYSIS.

⑨ Master's THESIS

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⑪ Dec 78

⑫ 137p.

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Larry J. Pulcher

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Abstract

The Test of Equality between Subsets of Coefficients in Two Regressions is developed and applied as a means to pre-screen variables from a regression model.

Some criterion for selection of variables are discussed and some existing regression packages are applied to data on characteristics of avionics equipment for comparison purposes.

# CRITERION FOR SELECTION OF VARIABLES IN A REGRESSION ANALYSIS

## I. Introduction

### Previous Results

At the request of the Systems Evaluation Branch of the Air Force Avionics Laboratory at Wright-Patterson AFB, the Westinghouse Electric Corporation performed a regression analysis of characteristics of Line Replaceable Avionics Units (LRU) in an attempt to model some of the equipment's logistics characteristics. Westinghouse identified 21 independent and six dependent variables. Seven of the 21 independent variables are purely indicators, identifying type of aircraft and general usage category of the equipment.

The regression was performed using the "Linear Least Squares Curve Fitting Program" (LLSCFP) developed by Daniel and Wood. In using LLSCFP, all 21 independent variables, as well as some terms containing the squares or the natural logarithms of variables are included simultaneously in the first regression. Then with the aid of statistics, plots, and tabular data arrangement, a subset collection of independent variables is chosen which best approximates the data.

### Scope of Present Study

There are several ways to arrive at the "best" subset of independent variables to include in a model. At one extreme, the model can consist of all possible variables. But this is not desirable for two reasons. First, it may be very expensive to gather and maintain a data base for a large number of variables, some of which may have little impact. But second and more important, when the purpose is to predict future costs, as it is in this study, a model that uses a large number of variables to fit the nuances of previous data may in fact have a higher prediction variance than a subset model (Ref 17:7). Important information could be lost in the myriad interrelationships that exist. For this reason the selection of the form and the variables of the regression equation becomes important.

The methods of selection of variables to be investigated in this study are: iterative techniques using the Statistical Package for the Social Sciences (SPSS) and BMD Biomedical Computer Programs (BMD), all possible regressions using the Leaps and Bounds technique developed by Furnival and Wilson, and the  $C_p$  statistic search using LLSCFP developed by Daniel and Wood.

The desired final outcome of this analysis is to provide the personnel at the Air Force Avionics Laboratory with a method of performing a quality regression without an extensive background in the technique, so that they can do work in-house which they previously



contracted out. For that reason the methods used in this work will rely on existing packages where possible.

### Theory of Linear Least Squares Regression

Assumptions. The first assumption made in the use of linear least squares regression is that the correct model has been chosen. If an incorrect form is used some values given by the equation will be biased.

The second assumption is that the data is typical of the true population about which the analysis is being performed.

The third assumption of the method is that the y observations are statistically uncorrelated and independent. If each y value is considered to be composed of a true and a random error value called  $\epsilon$ , then this assumption can be restated as: The expected value of the product of any two of the random components is zero.

Three other assumptions that are considered less important (Ref 8:8) are that all observations on y have the same unknown variance, that the levels of the independent variables are non-stochastic, and that the uncontrolled error is distributed normally.

Method of Least Squares. The general form of the linear least squares regression equation is

$$y = \alpha_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k + \epsilon \quad (1)$$

where y is the observed value of the dependent variable,  $x_i$  is the

observed value of the  $i^{\text{th}}$  independent variable, and  $\alpha_0$  and  $\beta_i$  are regression coefficients. While the linear squares method can treat only equations in the form of equation (1), there are non-linear equations which are intrinsically linear, such as  $y = \alpha_0 x_1^{\beta_1} x_2^{\beta_2}$ . By taking the natural log this equation becomes

$$\ln y = \ln \alpha_0 + \beta_1 \ln x_1 + \beta_2 \ln x_2$$

This is only one of a variety of intrinsically linear equations and methods for linearizing. Equations that are not intrinsically linear can not be handled using the linear least squares method. Non-linear least squares packages are available but will not be considered in this study.

If there are  $N$  independent observations of  $y_i$  and  $x_i$ , each observation can be written as

$$y_i = \alpha_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_K x_{iK} + \epsilon_i \quad (2)$$

where  $x_{ij}$  represents the  $i^{\text{th}}$  observation of the  $j^{\text{th}}$  variable. If the matrices

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_N \end{bmatrix} \quad (3)$$

$$X = \begin{bmatrix} 1 & x_{11} & x_{12} & \cdot & \cdot & \cdot & x_{1k} \\ 1 & x_{21} & x_{22} & \cdot & \cdot & \cdot & x_{2k} \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ 1 & x_{N1} & x_{N2} & \cdot & \cdot & \cdot & x_{Nk} \end{bmatrix} \quad (4)$$

$$\underline{\beta} = \begin{bmatrix} \alpha_0 \\ \beta_1 \\ \beta_2 \\ \cdot \\ \cdot \\ \beta_k \end{bmatrix} \quad (5)$$

$$\underline{\epsilon} = \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \cdot \\ \cdot \\ \cdot \\ \epsilon_N \end{bmatrix} \quad (6)$$

are defined, then all N equations can be written simultaneously as

$$\underline{y} = X \underline{\beta} + \underline{\epsilon} \quad (7)$$

The basis of the least squares method is that a straight line is generated through data points in such a way that the sum of the squared distances or errors between the line and the points is minimized. This sum of squared errors (SSE) can be written as  $\sum (\epsilon_i^2)$  and in matrix notation is

$$\sum (\epsilon_i)^2 = \underline{\epsilon}' \underline{\epsilon} \quad (8)$$



Substituting (7) into (8) yields

$$\underline{\epsilon}' \underline{\epsilon} = (\underline{y} - \underline{X} \underline{\beta})' (\underline{y} - \underline{X} \underline{\beta}) \quad (9)$$

which we would like to minimize. If classical optimization is performed on (9), an estimator of  $\underline{\beta}$

$$\underline{b} = (\underline{X}' \underline{X})^{-1} \underline{X}' \underline{y} \quad (10)$$

is found to minimize SSE.

The variation of the dependent variable measurements about their mean

$$(y_i - \bar{y})^2 \quad (11)$$

is called Total Sums of Squares (TSS). TSS can be decomposed to two components, the Sum of Squares Explained by the Regression (SSR) and the error unexplained by the regression (SSE) such that

$$SST = SSR + SSE \quad (12)$$

This partitioning gives rise to a measure of goodness of fit  $R^2_{yx}$  or commonly written as just  $R^2$  and called the Coefficient of Determination or the Multiple Correlation Coefficient Squared.  $R^2$  is defined as

$$R^2 = \frac{SSR}{SST} \quad (13)$$

and represents the fraction of the variability in the independent

variable explained by the regression equation. But it is known that the Multiple Correlation Coefficient calculated in this way is biased upward, always indicating a higher degree of correlation than actually exists in the true population. In order to correct for the bias,  $R^2$  is adjusted for degrees of freedom by the relationship

$$\bar{R}^2 = 1 - (1 - R^2) \left( \frac{N-1}{N-K-1} \right) \quad (14)$$

and called the Adjusted Multiple Correlation Coefficient. While  $\bar{R}^2$  is not entirely unbiased, it does exhibit less bias than  $R^2$ , and will be used in this analysis.

A third measure of goodness of fit is the  $C_p$  statistic derived by Mallows, and is based on total squared error. The total squared error can be considered to be made up of a squared bias plus a squared random error in  $y$  at each data point. If the total squared error is represented as

$$\sum_{i=1}^n (\nu_i - \eta_i)^2 + \sum_{i=1}^n \text{Var}(y_{ip}) \quad (15)$$

where  $\nu_i = \nu(X_{i1}, X_{i2}, \dots, X_{iN})$  is the expected value of  $y$  from an equation with true  $\beta$ s,  $\eta_i = b_0 + \sum_{j=1}^K b_j X_{ij}$  is the expected value of  $y$  from the equation of estimates of  $\beta$ s, and  $p$  is  $K + 1$ . The term  $\sum (\nu_i - \eta_i)^2$  can be represented by SSEB called the sum of squared errors bias. Also,  $\Gamma$  can be defined as the standardized total squared



error

$$\Gamma_p = \frac{SSEB_p}{\sigma^2} + \frac{1}{\sigma^2} \sum_{i=1}^N \text{Var}(y_{ip}) \quad (16)$$

It is known [Daniel and Wood (Ref 8:86)] that

$$\sum_{i=1}^N \text{Var}(y_{ip}) = p\sigma^2 \quad (17)$$

Combining (16) and (17) forms

$$\Gamma_p = \frac{SSEB_p}{\sigma^2} + p \quad (18)$$

The error sum of squares (SSE) is defined as

$$SSE = \sum_{i=1}^N (y_i - \hat{y}_i)^2 \quad (19)$$

and

$$E(SSE) = \sum_{i=1}^N (\nu_i - E(\hat{y}_i))^2 + (N-p)\sigma^2 \quad (20)$$

Because  $E(\hat{y}_i) = \eta_i$ , then

$$E(SSE) = \sum_{i=1}^N (\nu_i - \eta_i)^2 + (N-p)\sigma^2 \quad (21)$$

or

$$E(SSE) = SSEB_p = (N-p)\sigma^2 \quad (22)$$

Combining (18) and (22) gives

$$\Gamma_p = \frac{E(SSE_p)}{\sigma^2} - (N - 2p) \quad (23)$$

Now define  $C_p$  as an estimate of  $\Gamma_p$

$$C_p = \frac{SSE_p}{S^2} - (N - 2p) \quad (24)$$

where  $s^2$  is an estimate of  $\sigma^2$ . Note from (22) that when the correct model is used the bias,  $SSB_p$ , goes to zero and  $C_p$  goes to  $p$ .

Then the objective of the  $C_p$  search is to find the  $p$ -term equation which has a  $\frac{C_p}{p}$  value nearest one and therefore minimum bias. A drawback to the method is that it is sensitive to the  $S^2$  estimate of variance. The  $C_p$  obtained from two different models may not be comparable unless the same value of  $S^2$  was used in both. For this reason, the LLSCFP allows the option of using  $S^2$  from an entire set of input variables, from a subset of variables, or a user supplied value. It seems sensible, when a large number of models are being compared, to supply a constant value of  $S^2$  so the  $C_p$  values can be compared.

It should be noted that the form of the  $\Gamma_p$  and  $C_p$  equations (23) and (24) indicate the importance of eliminating unnecessary variables from a model. The removal of one variable has the capability to remove as much as two units from the standardized squared error.

## II. Selection of Variables

### Variables

Westinghouse collected data amounting to 63 points on 21 physical and usage characteristics and 6 logistics characteristics of line replaceable avionics units, with the intent to predict logistics characteristics from physical and usage characteristics. Sources used to collect the data were: existing Air Force Data Systems, site visits to Air Force logistics facilities, Westinghouse activities, published reports, and engineering analysis of LRUs. The following paragraphs describe briefly each of the variables. For a more in-depth description of the variables and the manner in which they were collected, see the Westinghouse report (Ref 22:18).

The first six independent variables are measures of physical characteristics. The Unit Price is measured in dollars per LRU. The Volume is measured in cubic feet. Weight is measured in pounds. Component Count is a measure of the number of electrical components of an LRU and does not include mechanical devices, connectors, or structure. Component Density is simply Component Count divided by Volume. While Westinghouse used Component Density in their analysis, it will not be used here, because the matrix of data would be singular in the log-linear model which will eventually be used. Power



Dissipation is measured in input watts minus output watts.

The next five independent variables, measures of component type, are in terms of percentages and are additive to unity. The variable names are descriptive, but if information on the way the measurements are determined is desired, the Westinghouse report (Ref 22:18) should be consulted. The variables are: Fraction Digital, Fraction Analog, Fraction Electromechanical, Fraction Power Supply, and Fraction Transmitter.

The twelfth variable, Fraction Solid State, is a measure of LRU technology, the percentage of components in the LRU that are solid state in nature.

The next data element collected concerned aircraft type and usage. Three types of aircraft were Fighter, Bomber, and Cargo. The three types of usage were Navigation, Sensory, and Communications. In both the Westinghouse and this study these parameters were used as indicators, but in different ways. Westinghouse coded them as follows:

Bomber	1	0
Cargo	0	1
Fighter	0	0
Sensory	1	0
Communications	0	1
Navigation	0	0

In addition, Westinghouse found interactions between certain of the two types useful in their analysis. These were coded as follows:

Bomber Sensory	1	0	0
Bomber Communications	0	1	0
Cargo Communications	0	0	1

In this study a different approach was taken to these variables. Interaction between all of the aircraft types and usages were considered as indicator variables and coded as follows:

Nav Fighter	1	0	0	0	0	0	0
Nav Bomber	0	1	0	0	0	0	0
Nav Cargo	0	0	1	0	0	0	0
Sensor Fighter	0	0	0	1	0	0	0
Sensor Bomber	0	0	0	0	1	0	0
Comm Fighter	0	0	0	0	0	1	0
Comm Bomber	0	0	0	0	0	0	1
Comm Cargo	0	0	0	0	0	0	0

The data base does not include any sensory equipment on Cargo aircraft, thus it is not used as a variable.

The final independent variable identified by Westinghouse is the Percentage of Failures Detected by Built-In-Test (BIT). This variable is intended to be a measure of the effectiveness of the Built-In-Test/Fault-Isolation-Test (BIT/FIT) capabilities of each LRU.

The six independent variables which Westinghouse identified are Maintenance Manhours/Operating Hour, Mean Time Between Failures, Mean Time Between Maintenance Actions, Logistics Support Cost/Operating Hour, Training Costs/Operating Hour, and Percentage Not Repairable This Station. The Operating Hours used to normalize is an estimate of the amount of time an LRU is turned on, data which is not recorded. Instead, Westinghouse used flying hours/year of the aircraft using an LRU multiplied by a scaling factor which they estimated

for each of the three types of aircraft. All of the maintenance and operating hours and cost data for these six variables are total yearly figures.

The previous analysis regressed the independent variables against each of the dependent variables individually. Because it is not the purpose of this study to dispute the earlier one, but rather to develop a method for the personnel of the Avionics Laboratory to perform the regression, only one of the dependent variables was chosen for illustration. Logistics Support Cost/Operating Hour was chosen because the Westinghouse report recommended that further study be conducted on it.

Table I contains the variables and the abbreviations used in this and the previous report. An asterisk indicates that the variable was not used in the study indicated. Table II contains the equation found by Westinghouse.

### Models

It is well known that the selection of the model is very important to the goodness of fit of a regression and its predictive ability. In light of this, a balance was sought that would yield a model with elements of both simplicity and goodness of fit.

The first model considered was the simplest form of multiple regression equation, namely



Table I  
List of Variables

Name	Abbreviations	
	Westinghouse	This Report
Unit Price	UP	UP
Volume	V	V
Weight	W	W
Component Count	CC	CC
Component Density	CD	*
Power Dissipation	PD	PD
Fraction Solid State	FSS	%SS
Fraction Digital	FDI	%DIG
Fraction Analog	FAN	%AN
Fraction Electromechanical	FEM	%EM
Fraction Power Supply	FPS	%PS
Fraction Transmitter	FXR	%XMTR
Bomber	IBOM	*
Cargo	ICAR	*
Sensory	ISEN	*
Communications	ICOM	*
Navigation-Fighter	*	NF
Navigation-Bomber	*	NB
Navigation-Cargo	*	NC

Table I (cont'd)

Name	Abbreviations	
	Westinghouse	This Report
Sensory-Fighter	*	SF
Sensory-Bomber	BOMSEN	SB
Communications-Fighter	*	CF
Communications-Bomber	BOMCOM	CB
Communications-Cargo	CARCOM	COMMC
Fraction BIT/FIT	BIT/FIT	BF
Logistics Support Cost/ Operating Hour	LSC/OH	LSC/OH
Maintenance Manhours/ Operating Hour	MMH/OH	*
Mean Time Between Failures	MTBF	*
Mean Time Between Maintenance Actions	MTBMA	*
Training Cost/Operating Hour	TRAIN/OH	*
Not Repairable This Station	NRTS	*
* Not used in the analysis.		



Table II

Logistics Support Cost/Operating Hour Equation  
Generated by Westinghouse

$$\ln (\text{LSC/OH}) = \alpha_0 + \sum_{i=1}^{21} b_i X_i$$

$$R^2 = .8916$$

$$R^2 = .9283$$

$$F\text{-value} = 25.3$$

i	$b_i$	$X_i$	Partial F
0	-8.15108		
1	3.86111	(IBOM - .286)	36.0
2	3.66533	(ICAR - .270)	31.4
3	$-4.85271 \times 10^{-1}$	(ISEN - .254)	3.6
4	-2.56663	(IBOM - .286) (ISEN - .254)	37.2
5	-1.66262	(IBOM - .286) (ICOM - .206)	12.2
6	$-7.67253 \times 10^{-1}$	(ICAR - .270) (ICOM - .206)	3.2
7	$1.27356 \times 10^{-2}$	FPS	6.8
8	$2.25967 \times 10^{-2}$	(FAN - 63.3)	36.0
9	$-7.42999 \times 10^{-3}$	(FSS - 61.1)	9.0
10	2.38503	(UP - 1.64)	27.0
11	$-9.20384 \times 10^{-11}$	(UP - 133606.3) <sup>2</sup>	25.0
12	$-1.52864 \times 10^{-4}$	(W - 64.3) <sup>2</sup>	8.4
13	$-1.07105 \times 10^{-3}$	(FAN - 48.9) <sup>2</sup>	33.6
14	$1.20418 \times 10^{-3}$	(FEM - 47.0) <sup>2</sup>	33.6
15	$7.10025 \times 10^{-4}$	(FXR - 40.2) <sup>2</sup>	10.9
16	$-1.61651 \times 10^{-4}$	(FSS - 51.9) <sup>2</sup>	2.2

Table II (cont'd)

i	$b_i$	$X_i$	Partial F
17	$-1.11568 \times 10^{-6}$	$(PD - 722)^2$	7.3
18	5.00996	$(UP - 1.68)^2$	42.2
19	$1.70042 \times 10^{-3}$	$(BF - 27.3)^2$	13.0
20	$4.60293 \times 10^{-1}$	LN(UP)	31.4
21	$2.35583 \times 10^{-1}$	LN(V)	4.8

$$y = \alpha_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k + \epsilon \quad (1)$$

When this model was used to form a prediction equation on the original data using SPSS, it did not provide satisfactory goodness of fit, having a Correlation Coefficient Squared of approximately 0.45, indicating that the values predicted by the model were poorly matched to the actual values. The model fell far short of the  $R^2$  of approximately 0.93 achieved by Westinghouse.

A model was desired which would allow more possibility of interactions and not be restricted to linearity of variables. At the same time it was noted that the five measures of component type, namely %DIG, %AN, %EM, %PS, %XMTR, along with the Fraction of Malfunctions Detected by BIT/FIT, BF, could all be converted to indicator form with little loss of information, as almost all of the measurements were near 0% or 100%. This meant that the variable list now

consisted of 13 indicator or dummy variables and 6 ordinary independent variables.

The model settled on was one of considerable complexity but offered many possibilities for interactions between variables. The Product of Powers model is of the form

$$y = e^{(\alpha_0 + \sum_i \alpha_i D_i)} \prod_j x_j^{(\beta_{j0} + \sum_i \beta_{ji} D_i)} \quad (25)$$

where  $\alpha$ ,  $\beta$ , and  $x$  have the same meaning as described in the previous model, and  $D_i$  is indicator or dummy variable  $i$  and  $j$  is the index of ordinary variable  $x$ . Because this model is not linear, least squares regression can not be used on it without a transformation. If the natural logarithm is taken on both sides of (25), the resulting equation is

$$\begin{aligned} \ln y = & \alpha_0 + \sum_{i=1}^{13} \alpha_i D_i + \sum_{j=1}^6 \beta_{j0} \ln x_j \\ & + \sum_{j=1}^6 \sum_{i=1}^{13} \beta_{ji} D_i \ln x_j \end{aligned} \quad (26)$$

Linear least squares regression can be applied to the model in (26), but there are now 97 possible variables. Because there are only 63 data points, selection of variables has become of paramount importance. And because only one of the packages used in this study, SPSS, is capable of handling this number of terms, some method was needed to



eliminate terms prior to regression. That is where a test of equality of regression populations was useful.

### Test of Equality of Regression Populations

The coefficients and constant terms,  $\beta$ s and  $\alpha$ s in (26) could be written in vector form as

$$\underline{\beta} = \begin{bmatrix} \alpha_0 + \alpha_1 + \dots + \alpha_{13} \\ \beta_{10} + \beta_{11} + \dots + \beta_{113} \\ \beta_{20} + \beta_{21} + \dots + \beta_{213} \\ \vdots \\ \beta_{60} + \beta_{61} + \dots + \beta_{613} \end{bmatrix} \quad (27)$$

by noting that  $\ln x_j$  is a common term in the second and third term of (26) and that  $D_i = 0$  or  $1$ . If only one indicator is considered at a time and all others are considered constant, (27) becomes

$$\underline{\beta_b} = \begin{bmatrix} a_0 + a_i \\ \beta_{10} + \beta_{1i} \\ \beta_{20} + \beta_{2i} \\ \vdots \\ \beta_{60} + \beta_{6i} \end{bmatrix} \quad (28)$$

Equation (28) can be divided into two subsets depending on whether  $D_i = 0$  or  $1$ .

$$\underline{\beta}_a = \begin{bmatrix} \alpha_0 \\ \beta_1 0 \\ \beta_2 0 \\ \cdot \\ \cdot \\ \beta_6 0 \end{bmatrix} \quad (29)$$

if  $D_i = 0$ . If  $D_i = 1$ , then equation (28) is the case.

If it could be shown that equation (28) and (29) are not significantly different, then (29) would be used and the interaction terms between that  $D_i$  and all  $x_j$  as well as that  $\alpha_i D_i$  would not be included in the model. Further, if it could be shown that  $\underline{\beta}_a = \underline{\beta}_b$  where

$$\underline{\beta}_a = \begin{bmatrix} \alpha_0 + \alpha_i \\ \beta_1 0 + \beta_1 i \\ \beta_2 0 \\ \beta_3 0 + \beta_3 i \\ \cdot \\ \cdot \\ \beta_6 0 + \beta_6 i \end{bmatrix} \quad (30)$$

and  $\underline{\beta}_b =$  equation (28), then the individual interaction between  $x_2$  and that  $D_i$  would not be needed in the model. Any of the  $\alpha_i$  or  $\beta_{ji}$  terms could be so tested either singly as above or in combinations.

Chow describes such a test (Ref 7:599) and calls it a Test of Equality Between Subsets of Coefficients in Two Regressions. The same technique was described by Fisher (Ref 12:364).

Under the alternative hypothesis  $\underline{\beta}_a \neq \underline{\beta}_b$  the model becomes

$$y_1 = X_1 \beta_1 + \epsilon_1 = Z_1 \gamma_1 + W_1 \delta_1 + \epsilon_1 \quad (31)$$

$$y_2 = X_2 \beta_2 + \epsilon_2 = Z_2 \gamma_2 + W_2 \delta_2 + \epsilon_2 \quad (32)$$

or

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} Z_1 & 0 & W_1 & 0 \\ 0 & Z_2 & 0 & W_2 \end{bmatrix} \begin{bmatrix} \gamma_1 \\ \gamma_2 \\ \delta_1 \\ \delta_2 \end{bmatrix} + \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \end{bmatrix} \quad (33)$$

where  $Z_1$ ,  $Z_2$ ,  $W_1$ , and  $W_2$  are submatrices of the X matrix of data

$$X = \begin{bmatrix} 1 & X_{11} & X_{12} & \dots & X_{1K} \\ 1 & X_{21} & X_{22} & \dots & X_{2K} \\ \dots & & & & \cdot \\ \dots & & & & \cdot \\ \dots & & & & \cdot \\ 1 & X_{N1} & \dots & & X_{NK} \end{bmatrix} \quad (4)$$

The matrix  $Z_1$  contains those elements of X for the variables which are being tested and in which the  $D_i$  being considered is equal to zero.  $Z_2$  contains those elements of X for the variables which are being tested and in which the  $D_i$  being considered is equal to one.  $W_1$  contains those elements of X for the variables which are not being tested and in which the  $D_i$  equals zero.  $W_2$  contains the remainder of X, those elements for which the variable is not being tested and  $D_i$  equals one. The vector  $\gamma_1$  contains the regression coefficients which are being tested, assuming  $D_i$  equals zero, such as  $\beta_{20}$  in (30). The



vector  $\gamma_2$  contains the coefficients being tested assuming  $D_i$  equals one such as  $\beta_{20} + \beta_{2i}$ . The vector  $\delta_1$  contains the regression coefficients not being tested assuming  $D_i$  equals zero. The vector  $\delta_2$  contains the coefficients not being tested assuming  $D_i$  equals one. For example, if it is desired to test whether

$$\begin{bmatrix} a_0 + a_1 \\ \beta_1 0 + \beta_1 1 \\ \beta_2 0 \\ \beta_3 0 + \beta_3 1 \\ \beta_4 0 + \beta_4 1 \\ \beta_5 0 + \beta_5 1 \\ \beta_6 0 + \beta_6 1 \end{bmatrix} = \begin{bmatrix} a_0 + a_1 \\ \beta_1 0 + \beta_1 1 \\ \beta_2 0 + \beta_2 1 \\ \beta_3 0 + \beta_3 1 \\ \beta_4 0 + \beta_4 1 \\ \beta_5 0 + \beta_5 1 \\ \beta_6 0 + \beta_6 1 \end{bmatrix} \quad (34)$$

then

$$Z_1 = \begin{bmatrix} X_{12} \\ X_{22} \\ \cdot \\ \cdot \\ \cdot \\ X_{n2} \end{bmatrix} \quad (35)$$

and

$$Z_2 = \begin{bmatrix} X_{n+12} \\ X_{n+22} \\ \cdot \\ \cdot \\ \cdot \\ X_{n+m2} \end{bmatrix} \quad (36)$$

Assuming that  $D_i$  equals zero for the first  $n$  observations and one for the next  $m$  observation.

$$W_1 = \begin{bmatrix} 1 & x_{11} & x_{13} & x_{14} & x_{15} & x_{16} \\ 1 & x_{21} & x_{23} & x_{24} & x_{25} & x_{26} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & x_{n1} & x_{n3} & x_{n4} & x_{n5} & x_{n6} \end{bmatrix} \quad (37)$$

and

$$W_2 = \begin{bmatrix} 1 & x_{n+11} & x_{n+13} & x_{n+14} & x_{n+15} & x_{n+16} \\ 1 & x_{n+21} & x_{n+23} & x_{n+24} & x_{n+25} & x_{n+26} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & x_{n+m1} & x_{n+m3} & x_{n+m4} & x_{n+m5} & x_{n+m6} \end{bmatrix} \quad (38)$$

The vectors would be

$$\gamma_1 = \begin{bmatrix} \beta_2 & 0 \end{bmatrix} \quad (39)$$

$$\gamma_2 = \begin{bmatrix} \beta_2 & 0 \end{bmatrix} + \begin{bmatrix} \beta_2 & 1 \end{bmatrix} \quad (40)$$

$$\delta_1 = \begin{bmatrix} a_0 \\ \beta_1 & 0 \\ \beta_3 & 0 \\ \beta_4 & 0 \\ \beta_5 & 0 \\ \beta_6 & 0 \end{bmatrix} \quad (41)$$



$$\delta_2 = \begin{bmatrix} a_0 + a_1 \\ \beta_1 0 + \beta_1 1 \\ \beta_3 0 + \beta_3 1 \\ \beta_4 0 + \beta_4 1 \\ \beta_5 0 + \beta_5 1 \\ \beta_6 0 + \beta_6 1 \end{bmatrix} \quad (42)$$

Under the null hypothesis  $\beta_a = \beta_b$ , equation (34), the model becomes

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} Z_1 W_1 0 \\ Z_2 0 W_2 \end{bmatrix} \begin{bmatrix} \gamma \\ \delta_1 \\ \delta_2 \end{bmatrix} + \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \end{bmatrix} \quad (43)$$

where  $Z_1$ ,  $Z_2$ ,  $W_1$ ,  $W_2$ ,  $\delta_1$ , and  $\delta_2$  are the same as in (33) and contains the regression coefficients of the variables being tested assuming  $D_i$  equals zero. In the example being used here,

$$\gamma = \beta_2 0 \quad (44)$$

Under the null hypothesis, the least squares estimators of  $\gamma$ ,  $\delta_1$ , and  $\delta_2$  are

$$\begin{bmatrix} c_0 \\ d_1 0 \\ d_2 0 \end{bmatrix} = \begin{bmatrix} Z_1' Z_1 + Z_2' Z_2 & Z_1' W_1 & Z_2' W_2 \\ W_1' Z_1 & W_1' W_1 & 0 \\ W_2' Z_2 & 0 & W_2' W_2 \end{bmatrix}^{-1} \begin{bmatrix} Z_1' Z_2 \\ W_1' 0 \\ 0 W_2' \end{bmatrix} \cdot \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \quad (45)$$

where  $y_1$  is a vector of observed  $y$  values for which the corresponding  $D_i$  equals zero, and  $y_2$  is a vector of observed  $y$  values for which the corresponding  $D_i$  equals one. Continuing the same example as

previously,

$$Y_1 = \begin{bmatrix} y_1 \\ y_2 \\ \cdot \\ \cdot \\ y_n \end{bmatrix} \quad (46)$$

and

$$Y_2 = \begin{bmatrix} y_{n+1} \\ y_{n+2} \\ \cdot \\ \cdot \\ y_{n+m} \end{bmatrix} \quad (47)$$

Under the alternative hypothesis,

$$\begin{bmatrix} c_1 \\ c_2 \\ d_1 \\ d_2 \end{bmatrix} = \begin{bmatrix} Z_1 Z_1 & 0 \\ 0 & Z_2 Z_2 \\ W_1 Z_1 & 0 \\ 0 & W_2 Z_2 \end{bmatrix} \begin{bmatrix} 0 & Z_1 W_1 & 0 \\ Z_2 Z_2 & 0 & Z_2 W_2 \\ 0 & W_1 W_1 & 0 \\ W_2 Z_2 & 0 & W_2 W_2 \end{bmatrix} \cdot \begin{bmatrix} Z_1 & 0 \\ 0 & Z_2 \\ W_1 & 0 \\ 0 & W_2 \end{bmatrix} \cdot \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \quad (48)$$

Once the estimated coefficients are found, an F-test can be used to test the null hypothesis. If  $m > p$ , where  $p$  is the length of the vectors in (34) the test is

$$F_{(q, m+n-2p)} = \frac{\|Z_1 c_1 + W_1 d_1 - Z_1 c_0 - W_1 d_0\|^2 + \|Z_2 c_2 + W_2 d_2 - Z_2 c_0 - W_2 d_0\|^2}{\|y_1 - Z_1 c_1 - W_1 d_1\|^2 + \|y_2 - Z_2 c_2 - W_2 d_2\|^2} \cdot \frac{m+n-2p}{q} \quad (49)$$

If  $p-q \leq m \leq p$  where  $q$  is the number of variables being tested or the length of vector  $\gamma_1$  in (39), the test is

$$F_{(m-p+q, n-p)} = \frac{\|Z_1c_1 + W_1d_1 - Z_1c_0 - W_1d_{10}\|^2 + \|y_2 - Z_2c_0 - W_2d_{20}\|^2}{\|y_1 - Z_1c_1 - W_1d_1\|^2} \cdot \frac{n-p}{m-p+q} \quad (50)$$

If the calculated  $F$  value is greater than the table  $F$  value at the desired level of confidence, then reject  $H_0$  and include the interaction term in the model. If the calculated  $F$  is lower than the table  $F$ , accept  $H_0$  and do not include the interaction term in the model. If  $m < p-q$ , the test can not be performed in which case the null hypothesis has not been rejected and the variables are not included in the model.

In doing this study, the Chow test was performed on each  $\beta_{j0} + \beta_{ji}$  and each  $\alpha_0 + \alpha_i$  term individually for each combination of  $j$  and  $i$ . Thus the test had the potential to eliminate any of 78  $\beta_{ji}$  and 13  $\alpha_i$  terms of the 97 possible in the product of powers model used.

To perform the calculations for these tests three programs were written. The first program called YSEPR simply separated the observed values of  $y$  into the vectors  $y_1$  and  $y_2$  for each of the 13 dummy variables. A listing of the program is contained in the appendix on Figure 4.

The second program, called SUMS, was written to calculate



all of the combinations of sums of variables, sums of squared variables, and sums of cross products of variables grouped as in YSEPR by the value of each dummy variable. These sums are calculated because when the multiplications required in the large matrices in (45) and (48) are performed, the result would be some of those elements calculated by SUMS. Rather than perform the calculations repeatedly when only a small systematic change is needed for each test, they are calculated only once by SUMS and read when needed by the third program. A listing of SUMS is contained in the appendix on Figure 5.

The third program, called CHOW, performs the calculations in (34), (37), (38), and (39) and outputs the F values with the parameters necessary to test the hypotheses. A listing of CHOW is contained in the appendix on Figure 6. Table III contains the calculated F values generated by CHOW. Table IV contains the ranks of the F values generated.

In the tests performed here,  $p-q \leq m \leq p$  never occurred. Therefore, the degrees of freedom for all of the tests were  $q$  and  $m+n-2p$  or 1 and 49 where  $m+n = 63$  and  $p = 7$ .

The coefficients associated with dummy variables SB, CF, CB, and PS could not be tested because  $m$  was less than  $p-q$ . As a result, there were  $91-24 = 67$  variables tested using the Chow test. If the overall hypothesis that none of the  $\beta_{j0} + \beta_{ji}$  terms differ from  $\beta_{j0}$  or  $\alpha_0 + \alpha_i$  terms differ from  $\alpha_0$  is true, then the probability that at

least one of the individual hypotheses will be rejected would be less than  $\alpha^*$  where

$$\alpha^* = 1 - (1 - \alpha)^{67} \quad (51)$$

and is the level at which individual tests are conducted. Then

$$\alpha = 1 - (1 - \alpha^*)^{1/67} \quad (52)$$

If  $\alpha^* = 0.10$  is desired then the associated value of  $\alpha$  would be 0.00157. Using the tabled F-value for  $\alpha = 0.001$  assures that  $\alpha^*$  is less than 0.10. Values of  $\alpha^*$  less than 0.10 can not be tested at this time because available tables of F-values only go down to 0.001. When the calculated F is larger than  $F_{0.001, 1, 49} = 12.11$  the individual hypothesis is rejected and the interaction  $\beta_{ji}$  term can not be removed from the model on the basis of this test. Table III contains the calculated F-values for each of the tested terms. Variable  $\emptyset$  indicates the  $\alpha_{0i}$  term. The numbering system used in program CHOW is shown at the bottom of the table.

Only 13 terms had F-values lower than the critical F. These in addition to the 24 variables which could not be tested and therefore did not fail a test of the null hypothesis mean that 37 variables have been eliminated from the 97 started with, leaving 60 variables still in the model. While this appears disappointing at first glance, leaving more variables still in the model than LLSCFP or Leaps and Bounds can

Table III

Calculated F Values from CHOW Test

Dummy	Variable						6
	0*	1	2	3	4	5	
1	46.792	35.957	25.792	25.083	25.321	4.445	30.243
2	48.9942	31.165	27.368	17.834	37.558	2.207	13.168
3	47.188	35.020	20.261	42.983	.434	1.664	2.142
4	25.038	4.028	43.242	42.912	35.336	12.666	.138
5	NT	NT	NT	NT	NT	NT	NT
6	NT	NT	NT	NT	NT	NT	NT
7	NT	NT	NT	NT	NT	NT	NT
8	49.002	36.238	36.429	47.934	44.038	48.998	35.319
9	49.440	40.300	14.671	45.948	.735	7.368	.681
10	48.900	35.818	43.052	21.842	28.715	7.510	19.331
11	NT	NT	NT	NT	NT	NT	NT
12	54.984	42.913	45.636	35.172	15.782	17.991	15.147
13	46.648	39.791	22.918	39.312	11.510	15.345	11.583
Variable Names							
	1	2	3	4	5	6	
	UP	V	W	CC	%SS	PD	
Dummy Names							
1	2	3	4	5	6	7	8
9	10	11	12	13			
NF	NB	NC	SF	SB	CF	CB	DIG
AN	EM	PS	XMTR	BF			
NT indicates insufficient points in a subset to test.							
*Variable $\emptyset$ indicates the $a_i D_i$ term.							



handle, experimentation will be conducted in later chapters in which the rankings of the F-values will be used to try to preselect a set of variables for a model. The rankings of the F-values are shown in Table IV, with the lowest F-value being ranked 1. Also sub-optimized solutions, in which the 60 variables are broken into subsets to search for smaller subsets that can be combined, to form manageable sets, will be tried. Table V contains a list of the variables remaining in the model after the Chow test.

Table IV  
Rank of F-Values from CHOW Test

Dummy	Variable						
	0*	1	2	3	4	5	6
1	55	39	29	27	28	9	32
2	59	33	30	20	42	7	15
3	56	34	23	48	2	5	6
4	26	8	50	46	37	14	1
5	NT	NT	NT	NT	NT	NT	NT
6	NT	NT	NT	NT	NT	NT	NT
7	NT	NT	NT	NT	NT	NT	NT
8	61	40	41	57	51	60	36
9	62	45	16	53	4	10	3
10	58	38	49	24	31	11	22
11	NT	NT	NT	NT	NT	NT	NT
12	63	47	52	35	19	21	17
13	54	44	25	43	12	18	13
<p>NT indicates that there were insufficient points in one of the subsets to perform the test.</p> <p>*Variable <math>\emptyset</math> indicates the <math>a_i D_i</math> term.</p>							

Table V

Variable Remaining in the Model After the Chow Test

1 UP	21 NF*V	41 DIG*%SS
2 V	22 NF*W	42 DIG*PD
3 W	23 NF*CC	43 AN*UP
4 CC	24 NF*PD	44 AN*V
5 %SS	25 NB*UP	45 AN*W
6 PD	26 NB*V	46 EM*UP
7 NF	27 NB*W	47 EM*V
8 NB	28 NB*CC	48 EM*W
9 NC	29 NB*PD	49 EM*CC
10 SF	30 NC*UP	50 EM*PD
11 SB	31 NC*V	51 XMTR*UP
12 CF	32 NC*W	52 XMTR*V
13 CB	33 SF*V	53 XMTR*W
14 DIG	34 SF*W	54 XMTR*CC
15 AN	35 SF*CC	55 XMTR*%SS
16 EM	36 SF*%SS	56 XMTR*PD
17 PS	37 DIG*UP	57 BF*UP
18 XMTR	38 DIG*V	58 BF*V
19 BF	39 DIG*W	59 BF*W
20 NF*UP	40 DIG*CC	60 BF*%SS
		61 LSC/OH (Dependent Variable)



### III. Stepwise Regression

#### Mechanics

The three most commonly used iterative techniques for determining the proper variables in a regression are; backward elimination, forward selection, and stepwise regression.

All three of the above methods make use of the Partial F-test. In this test, the explained sums of squares (SSR) are decomposed into components attributable to each independent variable.

In the standard regression method of decomposition, each variable is treated as if it had been added to the regression equation in a separate step after all other variables had been included. These F-values are then used to determine the next variable to enter or leave the equation, depending on the type of iterative regression being performed. The F-value is given by

$$F = \frac{\text{SSR due to } x_i / 1}{\text{SSE} / (N - K - 1)} \quad (\text{Ref 20:336})$$
$$= \frac{r_{y(i.1,2,\dots,K)}^2 / 1}{1 - R_{y.1,2,\dots,K}^2 / (N - K - 1)} \quad (53)$$

The term  $r_{y(1.1,2,\dots,K)}^2$  is the part correlation indicating the relationship between the observed  $y$  and the residual of independent variable  $i$

from which the effects of the other independent variables have been removed.

If the hierarchical decomposition method is used instead of the standard regression method, the order of inclusion must be specified and is used to determine the order to enter variables rather than a partial F test. The variable included first, the one with the highest assigned inclusion level, is evaluated by the ratio,

$$F = \frac{r_{y1}^2/1}{(1 - R_{y.1,2,\dots,K}^2)(N-K-1)} \quad (54)$$

The second regression coefficient is tested by the ratio,

$$F = \frac{r_{y(2.1)}^2/1}{(1 - R_{y.1,2,\dots,K}^2)/(N-K-1)} \quad (55)$$

Ref 20:337)

$$= \frac{\text{incremental SS due to } x_2/1}{\text{SSE}/(N-K-1)}$$

Each successive variable to be included in a hierarchical fashion would be evaluated in the same manner as indicated above, with a squared part correlation of the form  $r_{y(i.1,2,\dots,i-1)}^2$ , where  $i$  is the variable being added, in the numerator. All of the ratios above should be compared to the tabled F distribution with 1 and  $(N-K-1)$  degrees of freedom.

The hierarchical method of regression is used when there is some basis for believing that some variables will explain more variance than others before the regression is accomplished. Because there were no prior feelings about the variables used in this analysis, the hierarchical method was not used.

In the background elimination method, a regression equation with all of the possible terms is the starting point. Then partial F-value is calculated for each variable. If the lowest partial F-value is less than some preselected F-value, the variable corresponding to the value is removed from consideration, and the procedure is repeated from calculation of F-values for the new equation. If at any iteration there are no variables that have an F-value low enough to cause removal, the equation is adopted as calculated.

The forward selection method operates by entering variables one at a time until a satisfactory equation is reached. The order of inclusion is determined by using partial correlation coefficients as a measure of relative importance of the variables not yet in the equation. At each step, the variable with the highest partial correlation coefficient, that is, the variable with the highest correlation to the dependent variable after allowing for the effect of previously included variables, is brought into the equation. This procedure is repeated until the partial F-value of the latest entered term is less than some preselected value; then the equation is adopted.



The stepwise regression method is a refinement of the forward selection method in that at every stage of the procedure, the variables included in the model in earlier stages are examined. Thus a variable which was entered at an earlier stage but has been rendered unimportant by its relationship to later entered variables will be detected and removed from the equation. Again, this is done by comparing the partial F-values to a preselected F-value to find any variables to be removed. In later iterations, the removed variable is treated the same as a variable that has never entered the equation. The criterion for inclusion into the stepwise model is the same as in the forward selection method.

While calculation of the partial F-values needed to use these regression methods would be too time consuming to perform manually, both of the regression packages used in this chapter, SPSS and BMD, provide them, making the two packages easy and convenient to use. Because stepwise regression incorporates the main features of backward elimination and forward selection, it is considered to be the best of the three methods (Ref 11:172). While it has been shown that the three methods do not always choose the same subset (Ref 17:9), the stepwise method is more likely to choose the best. For that reason, of the three methods, stepwise regression will be used here.

In using iterative regression method, two pitfalls should be kept in mind. First, no significance should be attached to the order in which

variables are entered or removed from the model. The first variable entered is not necessarily the most important one when other variables have been entered. The partial F-value of each variable must be considered for each variable after each step to determine relative significance. Secondly, there is no guarantee that any of the three methods will supply an optimal equation, because of the restriction of removing and entering one variable at a time.

It was previously stated that preselected values of the F distribution are used as criterion in including and deleting variables from the equation. In both the SPSS and BMD packages these values are called FIN and FOUT respectively. A problem arises in preselecting these values, as the degrees of freedom are not known exactly before a regression is run. The tabled F-value has degrees of freedom 1 and  $(N-K-1)$  where N is the number of data points and K is the number of independent variables in the equation. As the primary goal of an analyst is normally to maximize some measure of goodness-of-fit, such as R, while a secondary goal is to minimize K, K is not known exactly before the regression. Therefore, the values of FIN and FOUT must be determined from an estimated K. As a result some experimentation may be necessary to obtain the value of K which yields desired values for the goodness-of-fit measure.

An alternative is to use the default F-values for FIN and FOUT, 0.01 and 0.005, respectively. These values are designed to include

all variables and not remove any from the equation; in effect, a forced fit unless there are a large number of variables. In the course of the analysis, regressions were performed on identical data using varying values for FIN and FOUT. When values other than default were used, truncated versions of the same equation obtained from the default values were identified. It appears the simplest procedure may be to use default values of FIN and FOUT until the value of K is known, then use the table value of F at degrees of freedom 1 and (N-K-1) as FIN. FOUT should be slightly smaller than FIN, but other than that its selection is rather arbitrary.

#### The Packages

As would be expected when using packages that operate in the same way, SPSS and BMD yield the same answers and even employ the same format of output. Each prints at each step, the value of the multiple correlation coefficient, R, the standard error of estimate, an analysis of variance table, the coefficients of all variables currently in the equation, the standard error of the coefficients, and the partial F-value of all independent variables whether or not in the equation. In addition, SPSS provides  $R^2$  and  $\bar{R}^2$ . Both also list the value of calculated F to test the hypothesis



$$H_0: \begin{bmatrix} \alpha \\ \beta_1 \\ \beta_2 \\ \cdot \\ \cdot \\ \cdot \\ \beta_K \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \cdot \\ \cdot \\ \cdot \\ 0 \end{bmatrix} \quad (56)$$

Plots of residuals versus the sequence of cases in a file can be selected as options in both packages. Additionally, BMD will make plots of residuals versus all or selected independent variables as an option. Other options for both input and output are available and identified in the appropriate manuals (Ref 20:352) and (Ref 10:235). Examples of output are available in both manuals (Ref 20:360) and (Ref 10:249) and partial output prepared for this report are included in the appendix in Figures 7 and 8.

SPSS is capable of handling up to 100 independent variables per fit and BMD can handle up to 80. While both of the packages have the capability to compute transformations, the feature was not used. Instead, all interactions were computed using a short FORTRAN program and stored with the original variables on a permanent storage file to avoid recalculating them numerous times. Table VI indicates the location of all variables on the tape used for both stepwise regression and Leaps and Bounds.

### Numbers of the Variables on Permanent File Used in SPSS and Leaps and Bounds

40

## Results

First stepwise regression was used on the 60 variables remaining after the Chow test. Table VII provides a summary of the independent variable included by each step of the regression along with  $R^2$  and  $\bar{R}^2$  for the first 48 steps. It can be seen that step 25 shows a jump of almost 0.03 in  $R^2$  and more than 0.04 in  $\bar{R}^2$  over the previous step. Steps that follow provide only small increases in  $R^2$  and  $\bar{R}^2$ . The adjusted  $R^2$  hits a peak at step 45 but actually changes very little between steps 25 and 45. After step 45,  $\bar{R}^2$  shows a general downward trend except when a variable is removed. As a result the equation at step 25 was chosen as the best compromise between correlation and number of variables included. The equation has 23 variables, which is two more than the Westinghouse analysis, but it also provides an  $R^2$  of more than 0.02 higher and  $\bar{R}^2$  of about 0.03 higher than they achieved. Table VIII contains the coefficients and the partial F-values for each variable in the equation.

Next, as a comparison, a stepwise regression was performed with all 97 variables as inputs, regardless of results of the Chow test. The first six variables selected were the same as in the previous run, but from that point the two equations diverged. The regression with all 97 variables consistently had an  $R^2$  of 0.02 to 0.03 lower than the previous run for the same number of variables included. This could be taken to mean that the variables deleted by the Chow test were



Table VII

Sequence of Stepwise Regression on 60 Variables  
Remaining After Chow

Step	Variable	$R^2$	$\bar{R}^2$
1	W	.57312	.56612
2	AN*UP	.65870	.64733
3	SB	.72117	.70700
4	DIG*W	.76054	.74402
5	NB	.78460	.76571
6	XMTR*CC	.80354	.78249
7	DIG*V	.81802	.79486
8	BF*%SS	.82905	.80372
9	NB*V	.83850	.81107
10	AN	.84931	.82034
11	EM	.86207	.83232
12	PS	.87267	.84211
13	BF*W	.87661	.84387
14	SF	.88006	.84508
15	UP	.88449	.84822
16	%SS	.88799	.84903
17	XMTR*%SS	.89475	.85499
18	NB*W	.90005	.85917
19	AN(removed)	.90005	.86229
20	EM*V	.90507	.86624

Table VII (cont'd)

Step	Variable	$R^2$	$\bar{R}^2$
21	NB*UP	.90861	.86822
22	AN*W	.91318	.87184
23	DIG	.91814	.87621
24	NF*CC	.92491	.88361
25	NF*UP	.95212	.92388
26	NC	.95429	.92543
27	CC	.95678	.92757
28	DIG*CC	.95891	.92923
29	NF*PD	.96101	.93094
30	SF*%SS	.96283	.93223
31	SF*W	.96431	.93294
32	SF*CC	.96921	.94035
33	NF	.97111	.94221
34	BF*UP	.97226	.94268
35	XMTR*PD	.97351	.94220
36	PD	.97482	.94228
37	EM*CC	.97564	.94261
38	V	.97703	.94217
39	BF	.97821	.94106
40	EM*PD	.97886	.94056
41	CC(removed)	.97886	.94294

Table VII (cont'd)

Step	Variable	$R^2$	$\bar{R}^2$
42	CF	.97918	.94218
43	EM*UP	.97975	.94129
44	BF*W(removed)	.98027	.94373
45	NC*V	.98080	.94530
46	EM*W	.98189	.94415
47	XMTR*UP	.98221	.94227
48	XMTR	.98101	.94112

indeed not needed and are only clouding the issue when they are allowed in the set being considered.

Finally, an SPSS run was made on the 35 variables that had the highest F-values from the Chow test in an attempt to determine whether the rank of the F-values had any significance. Even with all 35 variables in the equation the  $R^2$  was only 0.88 and  $\bar{R}^2$  was only 0.73. This indicates that the variables with the highest F-values are not necessarily the most important in a regression. The ranks of F-values should not be given any significance.

#### Conclusion

Both SPSS and BMD are very easy to use and require little foreknowledge of the basis of regression. Therein also lies a danger that



Table VIII

Coefficients of the SPSS Regression Equation

$R^2 = 0.95212$ $\bar{R}^2 = 0.92388$ $F = 33.72$			
$\ln (LSC/OH) = \alpha_0 + \sum_i \alpha_i D_i + \sum_j \beta_j \ln x_j + \sum_j \sum_i \beta_{ji} D_i \ln x_j$			
Variable No.	Variable Name	Coefficient	Partial F
1	UP	0.402702	13.63
3	W	0.084548	0.10
5	%SS	0.412407	37.28
8	NB	11.320694	23.80
10	SF	-1.135445	17.68
11	SB	-1.457859	26.48
14	DIG	3.710527	7.25
16	EM	-2.950970	9.44
17	PS	-0.092716	0.09
20	NF*UP	0.322015	0.07
23	NF*CC	-0.568085	27.14
26	NB*UP	-0.729848	7.51
27	NB*V	-1.803242	9.46
28	NB*W	2.506829	12.27
63	DIG*V	-1.995969	18.20
64	DIG*W	3.034970	17.51
68	AN*UP	-0.272142	7.44

Table VIII (cont'd)

Variable No.	Variable Name	Coefficient	Partial F
70	AN*W	0.758240	8.11
75	EM*V	0.422377	8.71
89	XMTR*CC	0.294839	25.70
90	XMTR*%SS	-0.456146	24.86
94	BF*W	0.697895	25.90
96	BF*%SS	-0.642736	43.88
Constant		-5.315378	79.01

the packages will be misapplied and their output applied blindly. Both manuals (Ref 20:320) and (Ref 10:215) contain an introduction to regression which should be sufficient background for most cases. Of the two manuals, the SPSS is more detailed and easier to follow.

#### IV. All Possible Regressions

The number of possible subsets given  $K$  possible variables increases at the rate of  $2^K$  as  $K$  increases, and the number of calculations required to invert the moments matrix for each subset is of the order  $K^3$ . By taking advantage of the symmetry of moments matrices and the deletion of unneeded rows and columns as successive regressions are calculated, as well as storing moments matrices for later modification and use, the order of calculations for each subset is brought down to  $K^2$ . If only the regression coefficients, their variances, and the sum of squared errors, are wanted, the calculations required are of order  $K$  for each subset. If only the sum of squared errors is needed, the number of calculations is less than six per subset (Ref 13:500). But even this would mean  $9.5 \times 10^{29}$  calculations for 97 variables and  $6.4 \times 10^9$  calculations for only 30 variables just to calculate only the sums of squared errors for each possible regression. Clearly some way to eliminate some possibilities before calculating is needed to make the method feasible for problems of this magnitude.

This is what has been developed by Furnival and Wilson (Ref 13). Their technique, called Leaps and Bounds, is based on the fundamental fact that  $SSE(A) \leq SSE(B)$  where  $A$  is any set of independent variables



and B is any subset of A. It is impossible for any subset of A to have a lower error sum of squares than A. From this, we can use the SSE of A as a lower bound of the subsets (in the nomenclature of Furnival and Wilson, offspring of A). It is known that regressions of less variables than A that are not offspring of A have a lower SSE than A, then there is no need to investigate the offspring of A as possible optimal solutions. In this way, the number of regressions and calculations are reduced. The amount by which they are reduced is determined by how early in the branching process good lower bounds are found. If a good lower bound is located early, most of the regressions are eliminated from consideration before calculations are performed on them. It was noted in performing this study that the amount of execution time, and hence calculations, varied over a wide range for a given number of variables input. For instance, for the case of 29 independent variables, execution time ranged from 14 to 110 seconds, depending on the variables input.

To illustrate the method, an example from Furnival and Wilson (Ref 13:506) will be duplicated below in Figure 1. There are five independent variables numbered 1 to 5. Underlined variables are those that are going to be removed in offspring equations. Missing variables indicate those that have already been removed from the branch. The numbers in parentheses are the SSE values for the corresponding equation. For example, .1245 indicates the subset of variables containing

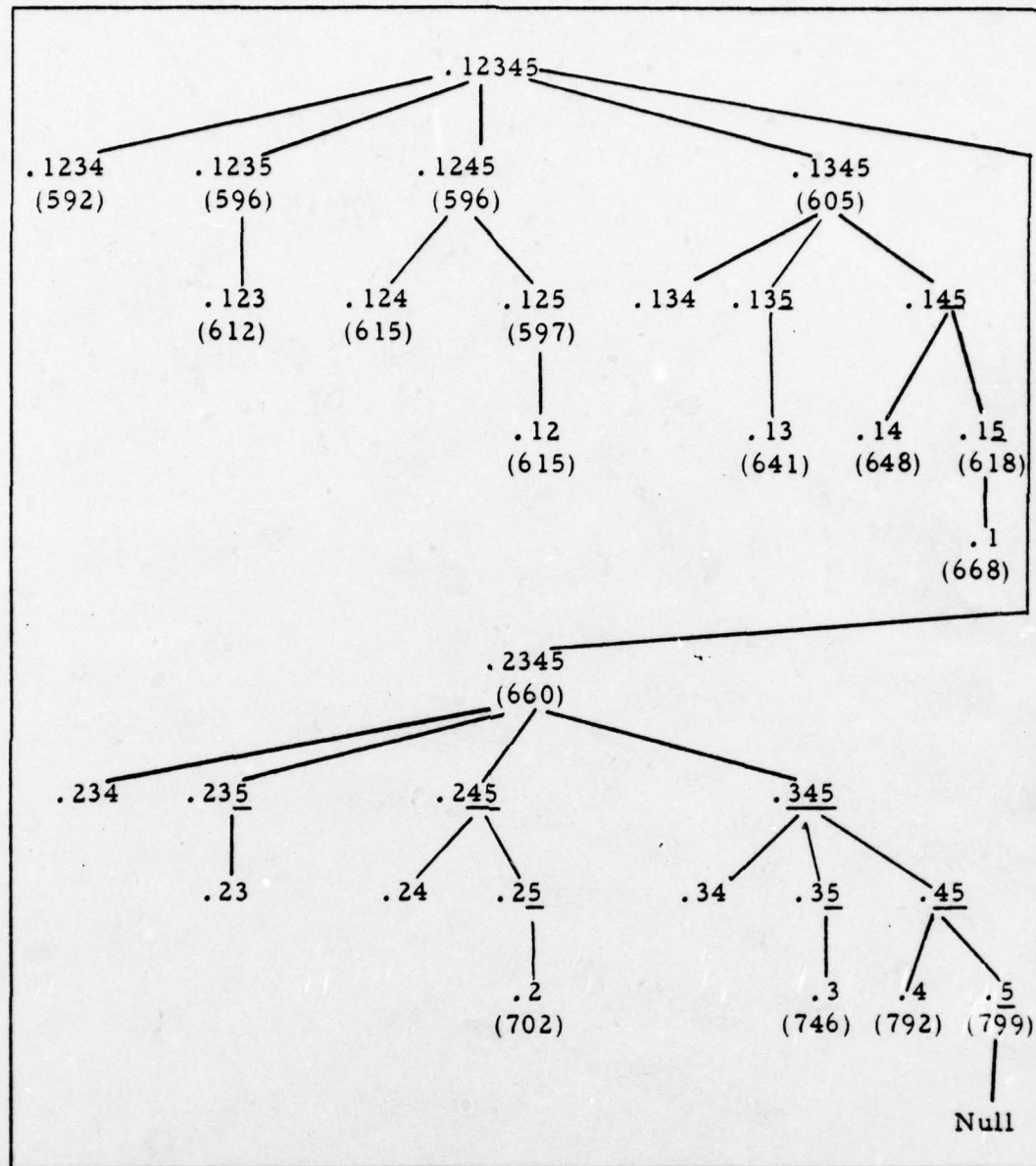


Figure 1. Tree Diagram

1, 2, 4, and 5, from which 4 and 5 will be removed in offspring branches.

There are various methods by which the tree can be traversed;

horizontally, vertically or a combination of those. The simplest for illustrative purposes is the horizontal method whereby equations of the same number of variables are considered together. The process begins by calculating the SSE for each of the four-variable equations. Then the SSE of equation .123 is calculated as 612. If 612 were lower than any of the SSEs from the four variable equations, no three-variable offspring of that four-variable equation would be considered or calculated. Note that equation .125 has an SSE of 597, which is lower than the SSE of 605 for equation .1345 or the SSE of 660 for equation .2345. Therefore, there is no need to calculate the SSE for equations .134, .135, .145, .234, .235, .245, or .345.

Next begin to evaluate the two-variable equations and compare them to the three and four-variable equations not already eliminated. Note that the SSE of equation .12, 615, is less than that for equation .2345. Therefore, there is no need to consider two-variable offspring of equation .2345. We need calculate the SSE only for equations .12, .13, .14, and .15.

Finally, begin to evaluate the one-variable models and compare them to the two, three, and four-variable models not already eliminated. If an SSE of a one-variable model is found to be less than that of a two, three, or four-variable equation, then their one-variable offspring need not be considered. In this case, all of the one-variable equations must be considered, and equation .1 is found to be the best



one-variable model.

In this example, only 17 values of SSE were calculated to guarantee that the best one, two, three, and four-variable equations were found. While the traverse used here is not as efficient as that used in the version of Leaps and Bounds in the International Mathematical and Statistical Libraries (IMSL), it is sufficient for illustrative purposes. The IMSL subroutine, called RLEAP, provides as output the  $m$  best of each of the 1 to  $K$ -variables subsets, where the user supplies the value of  $m$ . The output can be based on any or each of the criterion of  $R^2$ , adjusted  $R^2$  of  $C_p$ . The LLSCFP also uses a search technique based on  $C_p$ , but the version available for this study allowed only 29 variables to be input at a time and of these, only 12 at a time could be searched. This would cause the problem to become too fragmented for the number of variables that had to be considered. As a result, the LLSCFP was not used.

#### Application

It was found that the amount of execution time used by RLEAP grew very quickly above 20 variables. At 20 variables execution time was only a few seconds on the CDC 6600 series computer. But at 29 variables, the execution time could go as high as 138 seconds octal, at 35 variables over 1,000 seconds octal would be required. As a result 29 variables were settled on as a practice maximum input to RLEAP for this study.

Because there were 60 variables remaining from the Chow test, it was necessary to consider them in groups of no more than 29 and then to combine the optimum answers from each group into a new set. The new subset formed in this way was then run through RLEAP to pick the optimum subset from it. The result should then be a relatively small number of variables which can be used as a core to rotate the remaining variables through RLEAP again. The program used provided as output for each criterion the two best sets of variables for each equation size from 1 to the number of input variables, and the coefficients of the best subset. The procedure was begun by dividing the 60 variables into 3 groups of 20 and using RLEAP on each group. The three groups are shown in Table IX. Both  $\bar{R}^2$  and  $C_p$  statistics were used to search for best subsets. It was noted that the  $C_p$  statistic selected an equation of fewer variables than did the adjusted  $R^2$  criterion in most cases. In no case did the  $C_p$  criterion select an equation of more variables than did the adjusted  $R^2$  criterion. Because the purpose of this procedure is to find as small a subset as possible, the  $C_p$  criterion was used to select the best equation to be passed on to the next step, but the best adjusted  $R^2$  equation was also observed because that is a more easily interpreted statistic. The subsets of variables selected from the first three groups are shown in Table X along with the  $C_p$  for each subset. The three subsets selected were then combined into one set of 27 variables and again run through

Table IX

## Three Groups of Variables in First Run of RLEAP

Group 1	Group 2	Group 3
UP	NF·V	DIG·%SS
V	NF·W	DIG·PD
W	NF·CC	AN·UP
CC	NF·UP	AN·V
%SS	NB·V	EM·UP
NF	NB·W	EM·V
NB	NB·CC	EM·W
NC	NB·PD	EM·CC
SF	NC·UP	EM·PD
SB	NC·V	XMTR·UP
CF	NC·W	XMTR·V
CB	SF·V	XMTR·W
DIG	SF·W	XMTR·CC
AN	SF·CC	XMTR·%SS
EM	SF·%SS	XMTR·PD
PS	DIG·UP	BF·UP
XMTR	DIG·V	BF·V
BF	DIG·W	BF·W
NF·UP	DIG·CC	BF·%SS



Table X

Subset of Variables of Three Groups Selected by RLEAP

Group 1	Group 2	Group 3
UP	NF·W	DIG·%SS
W	NF·CC	DIG·PD
CC	NB·V	AN·W
NB	NB·W	EM·UP
SF	NC·V	EM·V
SB	NC·W	EM·W
DIG	SF·V	EM·PD
XMTR	SF·W	
	DIG·W	
Cp = 5.909	Cp = 2.958	Cp = 3.147

RLEAP to pick a smaller subset. The subset selected by this run is shown in Table XI below.

These 13 variables were then used as a core to add all of the remaining variables into three groups to give all variables a second chance to enter the equation. The new groups for this third run were formed by first including the 13 variables from the second run and then adding 15 or 16 of the 47 remaining variables until all are included in one group. Three groups of 28, 29, and 29 variables

Table XI

Subset Selected from 27 Variables Combined  
from Run 1 Based on Cp Criterion

UP	NF·CC	EM·
W	DIG·%SS	BF·W
SF	DIG·PD	BF·%SS
SB	AN·W	
DIG	EM·W	
$\bar{R}^2$ .83	Cp = 12.638	

resulted. When RLEAP was used on these three groups, the subsets selected by the Cp criterion shown in Table XII resulted. Asterisks indicate that the variable is one of the 13 identified by the previous run and input into all three groups. If the number of asterisks in a column is near 13, it indicates that the additional variables had little to offer in improving the model and that the model input from the previous run was relatively stable. If there are few asterisks, the additional variables had a lot to offer the model.

Until this point, the Cp statistic has been used as the criterion for choosing the best subset because the objective has been to eliminate a large number of variables quickly. But the subsets from Run 3 are small enough and the goodness-of-fit good enough for both  $\bar{R}^2$  and Cp criterion that both deserved further analysis. Table XIII

Table XII

Three Subsets Resulting from the Third Run  
Based on Cp Statistic

Group 1	Group 2	Group 3
W *	UP *	UP *
CC	W *	W *
%SS	SF *	SF *
NB	SB *	SB *
SF *	NF·W	NF·CC *
SB *	NF·CC *	DIG·CC
NF·UP	NF·PD	DIG·PD *
NF·CC *	NC·UP	AN·UP
DIG·PD *	NC·V	EM·W *
BF·W *	DIG·UP	EM·PD *
BF·%SS *	DIG·PD *	BF·W *
	AN·W *	BF·%SS *
	EM·W *	
	EM·PD *	
	BF·W *	
	BF·%SS *	
Cp = 6.802 $\bar{R}^2 = .8512$	Cp = 12.289 $\bar{R}^2 = .88$	Cp = 4.193 $\bar{R}^2 = .8436$
*Indicates the variable remains from the 13 from run 2.		



Table XIII

Three Subsets Resulting from the Third Run  
Based on  $\bar{R}^2$  Criterion

Group 1	Group 2	Group 3
UP *	UP *	UP *
W *	W *	W *
CC	SF *	SF *
%SS	SB *	SB *
NB	DIG *	DIG *
SF *	NF*W	NF*CC *
SB *	NF*CC *	DIG*V
DIG *	NF*PD	DIG*W
EM	NC*UP	DIG*%SS *
XMTR	NC*V	AN*W *
NF*UP	SF*CC	EM*W *
NF*CC *	DIG*UP	EM*PD *
DIG*%SS *	DIG*%SS *	BF*W *
DIG*PD *	DIG*PD *	BF*%SS *
EM*W *	AN*W *	
BF*W *	EM*W *	
BF*%SS *	EM*PD *	
	BF*W *	
	BF*%SS *	
$C_p = 9.632$ $R^2 = .8651$	$C_p = 12.52$ $R^2 = .8954$	$C_p = 6.0$ $R^2 = .8466$
*Indicates the variable remains from the 13 from Run 2.		

contains the three sets of variables selected in Run 3 by RLEAP using the  $\bar{R}^2$  criterion. The  $C_p$  values for the six sets in Tables XII or XIII can not be compared to each other because they have not been standardized as explained in Chapter I. But this will not cause a problem because the  $\bar{R}^2$  has a standard basis for all cases, and because the sets are going to be run through RLEAP again in various combinations as indicated in Figure 2. On the upper branch after Run 2 are the sets selected in Run 3 by the  $C_p$  statistic. Group 2 and Group 3 on this branch will be combined in Run 4 because they are similar. The set resulting from Run 4 will then be combined in Run 5 with the set in Group 1. By that point, the model should be stabilized if it is going to. The same scheme will be used on the  $\bar{R}^2$  branch of Run 3.

Table XIV shows the variables selected in each group and each run after the branch to Run 3. Figure 2 shows the branching with the results at each node in terms of number of variables,  $C_p$  value, and  $\bar{R}^2$ . It can be seen in Table XIV that in both the  $C_p$  and  $\bar{R}^2$  branches that Group 2 dominates the model. In the  $C_p$  branch the sets selected in Run 4 combining Groups 2 and 3 and Run 5 combining Group 1 and results from Run 4 are identical to Group 2. Apparently then Group 2 is the optimum subset of the variables in Run 3 from a  $C_p$  standpoint.

On the  $\bar{R}^2$  branch, Groups 2 and 3 combined in Run 4 to form a subset similar to Group 2. When Group 1 was combined in Run 5 with the set resulting from Run 4, the set selected was identical to the

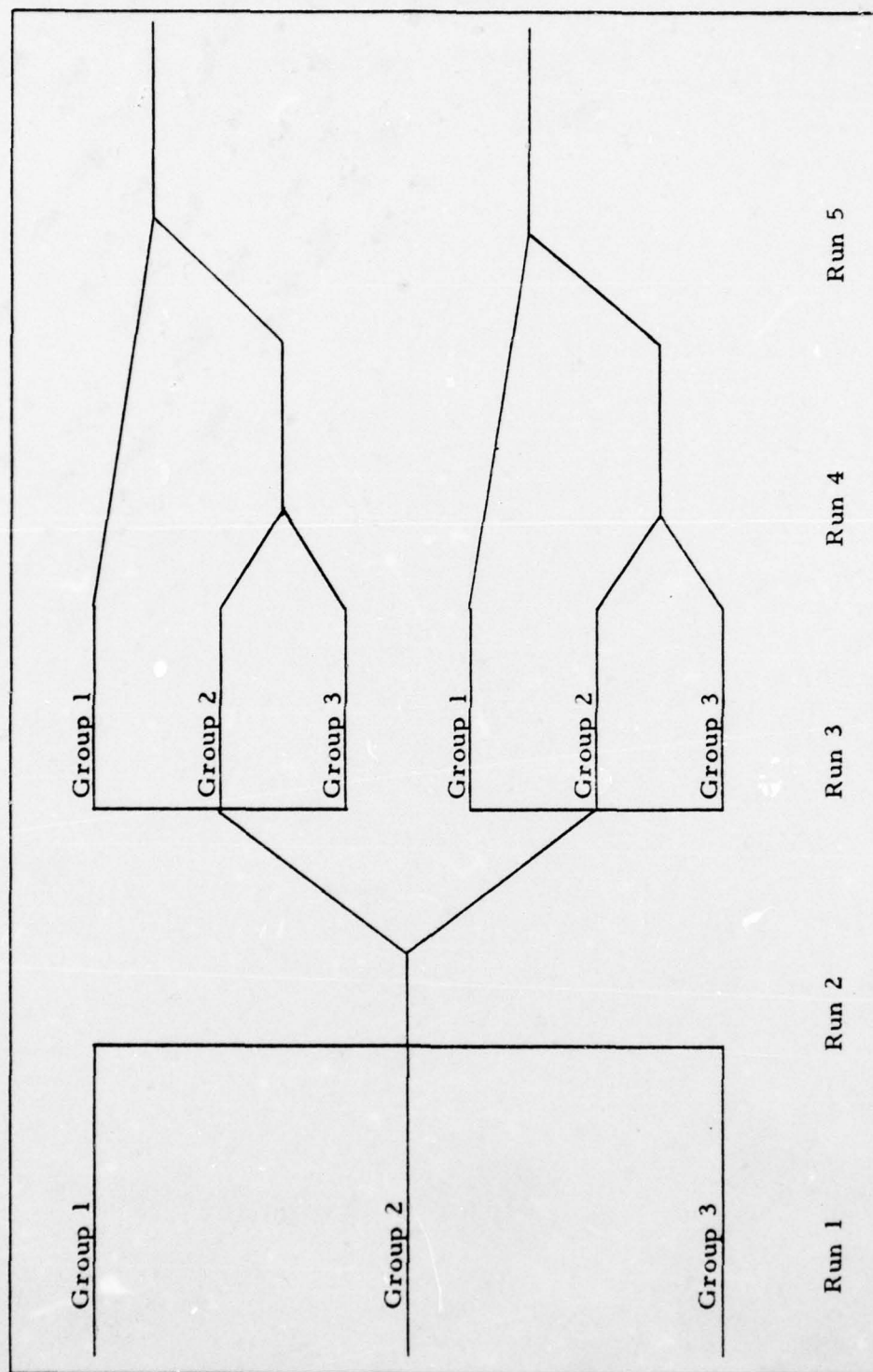


Figure 2. Sequence of Leaps and Bounds Runs



Table XIV

Variables Selected in Runs 3, 4, and 5

Variable	$C_p$ Criterion					$R^2$ Criterion				
	Run 3		Run 4		Run 5	Run 3		Run 4		Run 5
	Grp 1	Grp 2	Grp 3	Grp 2,3	Grp 1,2&3	Grp 1	Grp 2	Grp 3	Grp 2,3	Grp 1,2&3
UP		x	x	x	x	x	x	x	x	x
W	x	x	x	x	x	x	x	x	x	x
CC	x					x				
%SS	x					x				
NB	x					x				
SF	x	x	x	x	x	x	x	x	x	x
SB	x	x	x	x	x	x	x	x	x	x
DIG						x	x	x	x	x
EM						x				
XMTR						x				
NF*UP	x					x				
NF*W		x		x	x		x		x	x
NF*CC	x	x	x	x	x	x	x	x	x	x
NF*PD		x		x	x		x		x	x
NC*UP		x		x	x		x		x	x
NC*V		x		x	x		x		x	x
SF*CC							x		x	x
DIG*UP		x		x	x		x		x	x
DIG*V								x	x	x
DIG*W								x	x	x
DIG*CC		x								
DIG*%SS						x	x	x		
DIG*PD	x	x	x	x	x	x	x		x	x
AN*UP			x							
AN*W		x		x	x		x	x	x	x
EM*W		x	x	x	x	x	x	x	x	x
EM*PD		x	x	x	x		x	x	x	x
BF*W	x	x	x	x	x	x	x	x	x	x
BF*%SS	x	x	x	x	x	x	x	x	x	x

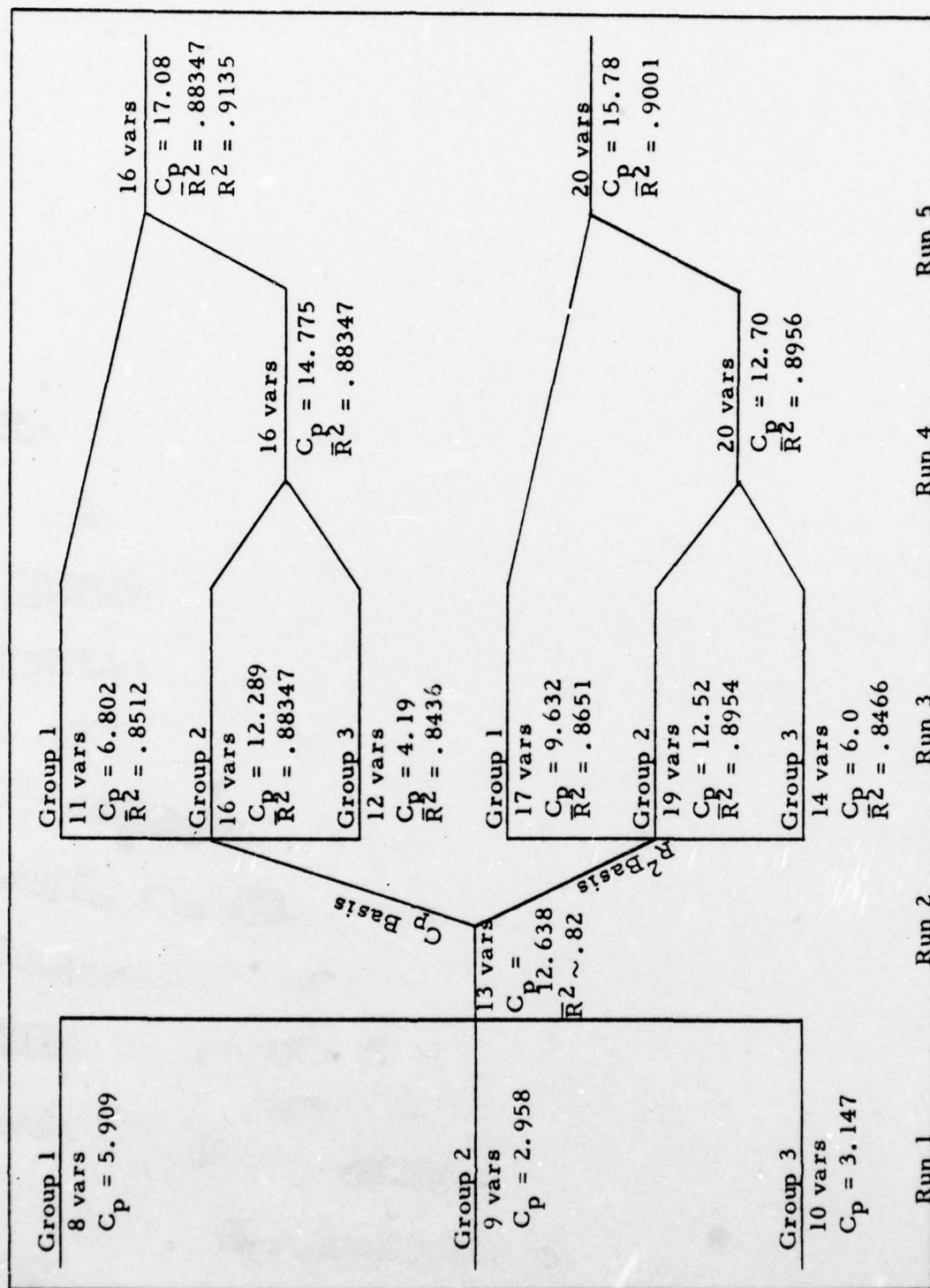


Figure 3. Results of Leaps and Bounds Runs

result from Run 4. Then the set from Run 4 must be the optimal subset of the variables in Run 3 from a  $\bar{R}^2$  standpoint.

The two equations found by the Leaps and Bounds method are described in Table XV, where the coefficients and partial F-values are given. It can be seen that the set generated using the  $C_p$  criterion is a subset of that generated by the  $\bar{R}^2$  method. The decision as to which equation is better will be deferred until Chapter V, when the SPSS generated equation will also be considered.

One thing to keep in mind is, as Aitken said (Ref 1:226), "There may be no one "best" equation, only a most appropriate one of several adequate equations."



Table XV

Equations Selected Using the All-Possible  
Regressions Method

$\ln (\text{LSC/OH}) = a_0 + \sum_i a_i D_i + \sum_j \beta_j \ln x_j + \sum_j \sum_i \beta_{ji} D_i \ln x_j$				
Variable	C <sub>p</sub> Criterion		R <sup>2</sup> Criterion	
	$R^2 = 0.9135$ $\bar{R}^2 = 0.88347$ $F = 31.21$		$R^2 = 0.9323$ $\bar{R}^2 = 0.9001$ $F = 29.25$	
	Coefficient	Partial F	Coefficient	Partial F
UP	0.245908	8.78	0.313871	14.52
W	0.384075	7.75	0.350494	6.86
SF	-1.061926	12.78	-2.878942	14.29
SB	-1.822390	30.26	-2.195891	39.06
DIG			4.381530	4.88
NF*W	-0.431742	31.61	-0.343076	2.10
NF*CC	-0.466254	13.70	-0.470354	15.84
NF*PD	0.738901	16.62	0.672722	14.59
NC*UP	0.285409	5.13	0.254284	4.04
NC*V	-0.334677	4.93	-0.292486	3.92
SF*CC			0.293229	6.30
DIG*UP	-0.584870	12.86	-0.950128	11.70
DIG*V			-0.971576	2.25
DIG*W			2.676919	4.93
DIG*PD	1.081951	15.97	0.553008	2.59
AN*W	0.309271	16.60	0.239272	9.98
EM*W	0.698175	13.89	0.705835	13.47
EM*PD	-0.555855	21.58	-0.545678	20.61
BF*W	0.866668	28.67	0.828916	27.04
BF*%SS	-0.701034	37.03	-0.706378	38.19
Constant	-3.855040	53.44	-4.091618	64.16
All other coefficients are zero.				

## V. Conclusions

### Results

Three equations have been found by stepwise regression and all possible regressions methods. They differ somewhat but have some terms in common and have similar performance characteristics. To aid in making a choice between the three, the variables contained in each and the pertinent statistics are summarized in Table XVI. The stepwise result has the higher correlation coefficients and overall F value, but it also has the most variables. Recall from Table VII that when the stepwise equation had 20 variables, as does the Leaps and Bounds  $\bar{R}^2$  equation, that the  $R^2$  was only 0.913 and  $R^2$  only 0.872. These values are somewhat lower than those for the Leaps and Bounds equations. This makes it appear that the Leaps and Bounds algorithm yields more efficient equations.

### Validation

Another way to distinguish between the two equations would be to use a validation procedure on all of them. There are six data points on LRUs which were not used in finding the models, which the equations will be tried against to see how well they predict. The data points are shown in Table XVII.

Table XVI

## Comparison of Three Regressions

Variable	SPSS	L&B Cp	L&B $\bar{R}^2$
UP	0.4027	0.2459	0.3139
W	0.0845	0.3841	0.3505
%SS	0.4124		
NB	11.3207		
SF	-1.1354	-1.0619	-2.8789
SB	-1.4578	-1.8224	-2.1959
DIG	3.7105		4.3815
EM	-2.9510		
PS	-0.0927		
NF·UP	0.3220		
NF·W		-0.4317	-0.3431
NF·CC	-0.5681	-0.4662	-0.4704
NF·PD		0.7389	0.6727
NB·UP	-0.7298		
NB·V	-1.8032		
NB·W	2.5068		
NC·UP		0.2854	0.2543
NC·V		-0.3347	-0.2925
SF·CC			0.2932
DIG·UP		-0.5849	-0.9501
DIG·V	-1.9960		-0.9718
DIG·W	3.0350		2.6769
DIG·PD		1.0820	0.5530
AN·UP	-0.2721		
AN·W	0.7582	0.3093	0.2893
EM·V	0.4224		
EM·W		0.6982	0.7058
EM·PD		-0.5558	-0.5457
XMTR·CC	0.2948		
XMTR·%SS	-0.4561		
BF·W	0.6979	0.8667	0.8289
BF·%SS	-0.6427	-0.7010	-0.7064
Constant	-5.3154	-3.8550	-4.0916
$R^2 = 0.95212$ $R^2 = 0.9135$ $R^2 = 0.9323$ $\bar{R}^2 = 0.9239$ $\bar{R}^2 = 0.8835$ $\bar{R}^2 = 0.9001$ $F = 33.72$ $F = 31.21$ $F = 29.25$			



Table XVII

## Data Points Used for Validation

	F4E	F4E	F4E	F4E	F111	F111
UP	8046	3398	9831	9910	5514	6650
V	585.2	323.7	562.7	776.5	1025.2	738.7
W	12.5	9.3	11.8	40.8	43.3	19.0
CC	878	58	209	73	1379	900
%SS	%87	%100	%77	%69	%78	%100
PD	58	75	24	1800	311.6	253.8
DIG	0	0	0	0	0	0
ANALOG	1	0	1	1	1	1
EM	1	0	1	1	0	0
PS	0	1	0	0	0	0
XMTR	0	0	0	0	1	1
BF	0	0	1	1	0	0
TYPE USE	SF	SF	SF	SF	NF	SF

While we could get point estimates plugging the data points into the regression equations we have generated, we would have no feel for the variability of our estimates. Prediction intervals could be generated that would give a level of confidence about the estimates. None of the packages mentioned up to the point provide the data needed

to calculate prediction intervals readily. But another simple statistics package, OMNITAB, written by the National Bureau of Standards, does. The formula used to calculate prediction intervals in this case is

$$PI = \hat{y} + t_{\frac{\alpha}{2}, N-K-1} \left( \sqrt{(SDPV)^2 + S^2} \right) \quad (57)$$

where  $\hat{y}$

$y$  is the point estimate

SDPV is the Standard Deviation of the Predicted Value from OMNITAB and

$S$  is the residual standard deviation.

All three of these values can be read directly from the OMNITAB output making the determination of prediction intervals quite easy. A listing of the OMNITAB output can be found in the Appendix in Figure 10. The OMNITAB package (Ref 18) will not perform any selection of variables, but once a subset has been decided on, it will provide useful information such as plots and the standard deviation measures needed to calculate the prediction intervals.

Table XVIII contains the prediction intervals of LSC/OH in terms of dollars/operating hour generated by each of the three equations for a 90% confidence interval.

Table XVIII

Generated by the Stepwise and the Leaps and Bounds Equations

	1	2	3	4	5	6
Actual	.589	.111	.513	.403	2.026	1.093
L&B Lower $C_p$	.0868	.0478	.0540	.0816	.4033	.1797
L&B $C_p$	.2350	.1273	.1504	.2432	1.1359	.4912
L&B Upper $C_p$	.6360	.3392	.4187	.7248	3.1992	1.3427
L&B Lower $\bar{R}$	.1253	.0328	.0477	.0643	.3786	.2376
L&B $\bar{R}$	.3314	.0866	.1252	.1916	1.0043	.6209
L&B Upper $\bar{R}$	.8765	.2284	.3288	.5709	2.664	1.6222
SPSS Lower	.0906	.1224	.0172	.2185	.2730	.1523
SPSS	.2089	.3068	.0431	.5218	.6658	.3619
SPSS Upper	.4818	.7689	.1079	1.247	1.6242	.8598
Westinghouse	.2952	.0909	.1313	.1598	1.2879	1.0418

The most noticeable result of comparing prediction intervals is that the SPSS equation generates 90% intervals that contain only one of the six actual values, while both Leaps and Bounds generated intervals contained five of the six actual values. On this basis, either of the Leaps and Bounds equations appears to be a better predictor than the SPSS equation. A closer look is required to choose between the two Leaps and Bounds equations, though.



First, the distance of the point estimate of each from the actual value was considered. In four cases, LRUs 1, 3, 4, and 5, the equation based on the  $C_p$  criterion had the point estimate nearest the actual value. Then the width of the 90% prediction interval was compared between the  $C_p$  and  $\bar{R}^2$  based criterion. Each criterion was found to have the narrower width of the prediction interval three times.

The  $C_p$  based equation appears to be the best predictor of those considered, based on generating the smallest error from actual values and its smaller size, 16 variables versus 20 or 23. This equation came closer than the Westinghouse equation to the actual values on two of the six cases, but Westinghouse did not generate prediction intervals to compare to. Two words of caution are necessary though. First, that the confidence levels used here can not be guaranteed to be known exactly. Refer to page 28 for the discussion of  $\alpha$  levels. Secondly, the six LRUs used for validation were all on fighter aircraft in usage category SF or NF and none were digital.

But with the information available, it appears that the personnel of the Air Force Avionics Laboratory could perform this same type of analysis using first the Chow test to pre-select variables for elimination, then applying the Leaps and Bounds method described in Chapter IV. The data collection is the most time consuming function and could still be contracted out. Computer analysis of the data as was done in this report could be done for less than \$100 of computer time using

packages already existing on the CDC system at Wright-Patterson AFB.

### Recommendations

At the present time, the Westinghouse Electric Corporation is in the process of enlarging the data base and performing another analysis. In progress reports to the Avionics Laboratory, they have mentioned the possibility of finding prediction intervals which could be compared to those generated for this report. In any case, the personnel of the Avionics Laboratory could use the enlarged data base to perform their own analysis as described in Chapter IV and if results are suitable, consider generating their own cost estimating relationships on future data.

All of the methods used in this report are based on minimizing the sum of squared errors. Other criterion for finding optimal subsets have been discussed in the literature in recent years but as yet not packaged for easy use. These include Mean Square Error of Prediction (Ref 2:469) and (Ref 6:46) and Average Estimated Variance (Ref 15:261). As these methods become proven, they should be considered for use on the data base, as they place more emphasis on prediction than do the criterion used in this study.

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APPENDIX

```

..
      PROGRAM YSEPR(INPUT,OUTPUT,TAPE1,TAPE2)
C
C      THIS PROGRAM FORMS THE Y1 AND Y2 MATRICES USED IN
C      PROGRAM CHOW
C
C      TAPE1 IS INPUT DATA, GUARIBLES, 13DUMMIES, AND
C      THE DEPENDENT VARIABLE IN THAT ORDER
C      FORMAT IS 4F20.14
C
C      TAPE2 IS OUTPUT
C
      DIMENSION X(63,20),Y1(63,13),Y2(63,13)
      DO 1 K=1,63
      READ(1,1000) (X(K,JJ),JJ=1,20)
1    CONTINUE
      DO 2 JJ=1,13
      IN=IN+1
      JD=JJ+6
      DO 3 K=1,63
      IF(X(K,JD).EQ.1.)GO TO 100
      IN=IN+1
      Y1(IN,JJ)=X(K,20)
      GO TO 3
100  IN=IN+1
      Y2(IN,JJ)=X(K,20)
3    CONTINUE
      PRINT 2002,IN
      WRITE(2,2002)IN
      PRINT 2000,(INC,JJ,Y1(INC,JJ),INC=1,IN)
      WRITE(2,2000) (INC,JJ,Y1(INC,JJ),INC=1,IN)
      PRINT 2003,IN
      PRINT 2001,(INC,JJ,Y2(INC,JJ),INC=1,IN)
      WRITE(2,2003)IN
      WRITE(2,2001)(INC,JJ,Y2(INC,JJ),INC=1,IN)
2    CONTINUE
1000 FORMAT(4F20.14)
2000 FORMAT(3(5X,"Y1(",2I2,")-",F11.8))
2001 FORMAT(3(5X,"Y2(",2I2,")-",F11.8))
2002 FORMAT(1X,"N=",I2)
2003 FORMAT(1X,"M=",I2)
      STOP
      END
SEOR
SEOR
SEOF
..

```

Figure 4. Program YSEPR



```

C      PROGRAM SUMS(INPUT,OUTPUT,TAPE1,TAPE2)
C
C      THIS PROGRAM CALCULATES ALL SUMS OF SQUARES AND
C      CROSSPRODUCTS NEEDED FOR PROGRAM CHOW
C
C      TAPE 1 IS INPUT DATA, 6VARIABLES AND 13DUMMIES AND
C      DEPENDENT VARIABLE
C      FORMAT IS 5(4F20.14)
C
C      TAPE2 IS OUTPUT
C
C      DIMENSION X(20),SS0(6,6,13),SS1(6,6,13),S0(6,13),S1(6,13)
C      DO 7001 JD=7,19
C      JJ=JD-6
C      DO 7002 J1=1,6
C      S0(J1,JJ)=0.
C      S1(J1,JJ)=0.
C      DO 7003 J2=1,6
C      SS0(J1,J2,JJ)=0.
C      SS1(J1,J2,JJ)=0.
7003 CONTINUE
7002 CONTINUE
7001 CONTINUE
1001 FORMAT(4F20.14/4F20.14/4F20.14/4F20.14/4F20.14)
PRINT 8000
8000 FORMAT('COMPLETED MODULE 1')
DO 1 J=1,63
READ(1,1001)(X(M),M=1,20)
DO 2 JD=7,19
JJ=JD-6
IF(X(JD).EQ.1.)GO TO 4
DO 3 J1=1,6
S0(J1,JJ)=S0(J1,JJ)+X(J1)
DO 3 J2=J1,6
SS0(J1,J2,JJ)=SS0(J1,J2,JJ)+(X(J1)*X(J2))
3 CONTINUE
GO TO 2
4 DO 5 J1=1,6
S1(J1,JJ)=S1(J1,JJ)+X(J1)
DO 5 J2=J1,6
SS1(J1,J2,JJ)=SS1(J1,J2,JJ)+(X(J1)*X(J2))
5 CONTINUE
2 CONTINUE
1 CONTINUE
PRINT 8001
8001 FORMAT('COMPLETED MODULE 2')
DO 6 JD=7,19
JJ=JD-6
DO 7 J1=1,5
JF=J1+1
DO 8 J2=JF,6

```

Figure 5. Program SUMS

```

      SS0(J2,J1,JJ)=SS0(J1,J2,JJ)
      SS1(J2,J1,JJ)=SS1(J1,J2,JJ)
      8 CONTINUE
      7 CONTINUE
      6 CONTINUE
      PRINT 8002
8002 FORMAT('MODULE 3 COMPLETED')
      DO 9 IDUM=1,2
      ITEM=IDUM-1
      DO 10 J1=1,6
      DO 11 J2=1,6
      IF(ITEM.EQ.1)GO TO 100
      PRINT 2001,(ITEM,J1,J2,JJ,SS0(J1,J2,JJ),JJ=1,13)
2001 FORMAT(3(1X,'SS',I1,'(',2I1,12,')',F16.10)/,3(1X,'SS',I1,'(',2I1,12,')',F16.10)/,3(1X,'SS',I1,'(',2I1,12,')',F16.10)/,3(1X,'SS',I1,'(',2I1,12,')',F16.10)/,3(1X,'SS',I1,'(',2I1,12,')',F16.10)/,3(1X,'SS',I1,'(',2I1,12,')',F16.10)/,3(1X,'SS',I1,'(',2I1,12,')',F16.10)/,3(1X,'SS',I1,'(',2I1,12,')',F16.10)
      WRITE(2,2001)(ITEM,J1,J2,JJ,SS0(J1,J2,JJ),JJ=1,13)
      IF(ITEM.EQ.0)GO TO 11
100 PRINT 2001,(ITEM,J1,J2,JJ,SS1(J1,J2,JJ),JJ=1,13)
      WRITE(2,2001)(ITEM,J1,J2,JJ,SS1(J1,J2,JJ),JJ=1,13)
      11 CONTINUE
      10 CONTINUE
      9 CONTINUE
      DO 12 IDUM=1,2
      ITEM=IDUM-1
      DO 13 J1=1,6
      IF(ITEM.EQ.1)GO TO 200
      PRINT 2002,(ITEM,J1,JJ,SS0(J1,JJ),JJ=1,13)
2002 FORMAT(3(1X,'S',I1,'(',I1,12,')',F16.10)/,3(1X,'S',I1,'(',I1,12,')',F16.10)/,3(1X,'S',I1,'(',I1,12,')',F16.10)/,3(1X,'S',I1,'(',I1,12,')',F16.10)/,3(1X,'S',I1,'(',I1,12,')',F16.10)/,3(1X,'S',I1,'(',I1,12,')',F16.10)
      WRITE(2,2002)(ITEM,J1,JJ,SS0(J1,JJ),JJ=1,13)
      IF(ITEM.EQ.0)GO TO 13
200 PRINT 2002,(ITEM,J1,JJ,SS1(J1,JJ),JJ=1,13)
      WRITE(2,2002)(ITEM,J1,JJ,SS1(J1,JJ),JJ=1,13)
      13 CONTINUE
      12 CONTINUE
      PRINT 8003
8003 FORMAT('COMPLETED MODULE 4')
      STOP
      END

```

PROGRAM CHOU(INPUT, OUTPUT, TAPE1, TAPE2, TAPE3, TAPE5-OUTPUT)

TAPE 1 IS OUTPUT FROM PROGRAM SUNS.  
TAPE 2 IS OUTPUT FROM PROGRAM YSEPARATOR.  
TAPE 3 IS DATA WITH 6 VARIABLES, 13 DUMMIES, AND 1 INDEPENDENT  
VARIABLES IN THAT ORDER FORMAT(5(420.14'))

CHOU IS DIMENSIONED TO HANDLE 63 OBSERVATIONS OF 6 INDEPENDENT,  
13 DUMMY, AND 1 DEPENDENT VARIABLES

REQUIRED INSL ROUTINES ARE:  
LINUZF  
UNULFF

DIMENSION Z1(63,1),U1(63,6),Z2(63,1),U2(63,6),V1(63,1),Y2(63,1)  
Z1TRAN(1,63),Z2TRAN(1,63),U1TRAN(6,63),U2TRAN(6,63),  
A1DATA(13,13),A2DATA(13,63),A3DATA(14,14),A4DATA(14,63),YMAT(63,1),  
A5DATA(13,13),A6DATA(14,14),S50(6,6,13),S51(6,6,13),S50(6,13),  
ZINTERO(13,63),ZINTERA(14,63),FINALO(13,1),FINALA(14,1),Z1C1(63,1),  
A1D1(63,1),Z1C0(63,1),U1D1O(63,1),Z2C2(63,1),U2D2(63,1),Z2C0(63,1),  
U2D2O(63,1),SUM1(63,1),SUM2(63,1),SUM3(63,1),SUM4(63,1),  
SUM5(63,1),C1(1),D1(6,1),C0(1),D1O(6,1),C2(1),D2(6,1),D2O(6,1),X(6  
3,19),S1(6,13),WORK(300)

THIS SECTION READS SUMS OF SQUARES FROM TAPE1

```
DO 1 I=1,6
DO 1 J=1,6
READ(1,50)(S50(I,J,J),JJ=1,13)
1 CONTINUE
DO 2 I=1,6
DO 2 J=1,6
READ(1,50)(S51(I,J,J),JJ=1,13)
2 CONTINUE
DO 3 I=1,6
READ(1,60)(S60(I,J,J),JJ=1,13)
3 CONTINUE
DO 4 I=1,6
READ(1,60)(S1(I,J,J),JJ=1,13)
4 CONTINUE
50 FORMAT(3I10X,F16.10)
60 FORMAT(3I8X,F16.10)
1000 FORMAT(3X,I2)
1001 FORMAT(3I14X,F11.8)
1010 FORMAT(1X,"JJ=",I2,2X,"UJJ=",I2,2X,"UJJ=",I2)
READ X DATA FROM TAPE 3
```

Figure 6. Program CHOU



```

DO 2000 K=1,63
  READ(3,3000)X(K,1),I=1,19)
  2000 CONTINUE
  3000 FORMAT(4F20.14/4F20.14/4F20.14/3F20.14)
C
C      JD IS THE DUMMY VARIABLE COUNTER
C
DO 9999 JD=7,19
  JJ=JD-6
  7474 OPT=1
  PROGRAM C10W
C
C      READ SEPARATED FROM TAPE 2 INTO V1 AND V2
C
  READ(2,1000)NJJ
  READ(2,1001)(V1(,INC,1),INC=1,NJJ)
  READ(2,1000)NJJ
  READ(2,1001)(V2(,INC,1),INC=1,NJJ)
C
C      JU IS THE INDEPENDENT VARIABLE INDICATOR
C
DO 9928 JU=1,7
  JULESS=JU-1
  IF(JU.GT.1)GO TO 3999
  C
  C      FILL AMAT0, AMATA TO TEST ALPHA(JD) TERM
  C
  AMAT0(1,1)=63
  AMATA(1,1)=NJJ
  AMATA(2,2)=NJJ
  DO 3001 I=1,6
    AMAT0(1,I)=50(1,JJ)
    AMAT0(1,I+1)=50(1,JJ)
    AMATA(1,I+2)=50(1,JJ)
    AMATA(1,I+2)=50(1,JJ)
    AMAT0(1,I+7)=51(1,JJ)
    AMAT0(1,I+7)=51(1,JJ)
    AMATA(1,I+8)=51(1,JJ)
    AMATA(1,I+8)=51(1,JJ)
    AMATA(2,I+8)=51(1,JJ)
  3001 CONTINUE
  DO 3002 I=1,6
    DO 3002 J=1,6
      AMAT0(I+1,J+1)=550(I,J,JJ)
      AMATA(I+2,J+2)=550(I,J,JJ)
      AMAT0(I+7,J+7)=551(I,J,JJ)
      AMATA(I+8,J+8)=551(I,J,JJ)
  3002 CONTINUE
C
C      FILL Z1, Z2, U1, AND U2 TO TEST ALPHA(JD) TERM
C
DO 3003 I=1,NJJ
  Z1(I,1)=1.0
  3003 CONTINUE
DO 3004 I=1,NJJ
  Z2(I,1)=1.0
  3004 CONTINUE
  IN=IN+6
  DO 3100 INC=1,63
    IF(X(,INC,JJ).EQ.1.0)GO TO 3150
    IN=IN+1
  3105

```

```

100 DO 3110 J-1.6
      UI(IN,J)-X(INC,J)
      3110 CONTINUE
      GO TO 3100
110 3150 IM-IN+1
      DO 3160 J-1.6
      U2(IN,J)-X(INC,J)
      3160 CONTINUE
      74/74 OPT-1
      PROGRAM CHOU
115 C
      C EXIT TO EVALUATION SECTION
      C
      3100 CONTINUE
      GO TO 8104
120 3999 IF(JV.GT.2)GO TO 4071
      C
      C FILL ANATA,ANATA TO TEST BETA(1,JD) TERM
      C
      4001 DO 4001 J-9.14
      ANATA(J,1)=0.0
      ANATA(1,J)=0.0
      4001 CONTINUE
      ANATA(1,1)=SS0(1,1,JJ)+SS1(1,1,JJ)
      ANATA(1,1)=SS0(1,1,JJ)
      ANATA(2,2)=SS1(1,1,JJ)
      ANATA(2,1)=SS0(1,1,JJ)
      ANATA(3,1)=SS0(1,1,JJ)
      DO 4000 I-3.7
      APAT0(I,1)=SS0(1,1-J,JJ)
      ANATA(1,1)=SS0(1,1-J,JJ)
      4000 CONTINUE
      ANATA(8,1)=SI(1,JJ)
      ANATA(9,2)=SI(1,JJ)
      DO 4010 I-9.13
      ANATA(1,1)=SS1(1,1-7,JJ)
      ANATA(1,2)=SS1(1,1-7,JJ)
      4010 CONTINUE
      APAT0(2,2)=NJJ
      ANATA(3,3)=NJJ
      DO 4020 I-3.7
      APAT0(1,2)=SS0(1,1-J,JJ)
      ANATA(1,3)=SS0(1,1-J,JJ)
      4020 CONTINUE
      DO 4030 J-3.7
      DO 4030 I-J.7
      ANATA(1,J)=SS0(1,1-J-1,JJ)
      ANATA(1,1)=SS0(1,1-J-1,JJ)
      4030 CONTINUE
      ANATA(8,8)=NJJ
      ANATA(9,9)=NJJ
      DO 4040 I-9.13
      ANATA(1,8)=SI(1,1-7,JJ)
      ANATA(1,9)=SI(1,1-7,JJ)
      4040 CONTINUE
      DO 4050 J-9.13
      DO 4050 I-J.13
      ANATA(1,J)=SS1(1,1-7,J,JJ)
      ANATA(1,1)=SS1(1,1-7,J,JJ)
      4050 CONTINUE

```

```

165      DO 4060 J-1,13
      DO 4060 I-1,13
      AMATA(J,I)=AMATA(I,J)
      CONTINUE
4060
170      DO 4070 J-1,14
      DO 4070 I-1,14
      AMATA(J,I)=AMATA(I,J)
      74/74 OPT-1
      PROGRAM CHOU
      4070 CONTINUE
      C
      C      EXIT TO FILL Z1, Z2, U1, UZ
      C
      C      GO TO 3050
      C
      C      FILL AMATO, AMATA TO TEST BETA(JV,JJ), JV=2 TO 6
      C
4071      AMATO(1,1)=SS0(JULESS,JJ)+SS1(JULESS,JULESS,JJ)
      AMATA(1,1)=SS0(JULESS,JJ)
      AMATA(2,2)=SS1(JULESS,JJ)
      AMATA(2,1)=SS0(JULESS,JJ)
      AMATA(3,1)=SS0(JULESS,JJ)
      DO 3010 I-1,5
      IF(I.GE.JULESS)GO TO 3005
      AMATO(I+2,1)=SS0(I,JULESS,JJ)
      AMATA(I+3,1)=SS0(I,JULESS,JJ)
      GO TO 3010
      3005      AMATO(I+2,1)=SS0(I+1,JULESS,JJ)
      AMATA(I+3,1)=SS0(I+1,JULESS,JJ)
      3010 CONTINUE
      AMATO(8,1)=S1(JULESS,JJ)
      AMATA(9,2)=S1(JULESS,JJ)
      DO 3020 I-1,5
      IF(I.GE.JULESS)GO TO 3015
      AMATO(I+8,1)=SS1(I,JULESS,JJ)
      AMATA(I+9,2)=SS1(I,JULESS,JJ)
      GO TO 3020
      3015      AMATO(I+8,1)=SS1(I+1,JULESS,JJ)
      AMATA(I+9,2)=SS1(I+1,JULESS,JJ)
      3020 CONTINUE
      DO 3021 I-1,12
      AMATO(I,1+1)=AMATO(I+1,1)
      3021 CONTINUE
      DO 3022 I-1,6
      AMATA(I,1+2)=AMATA(I+2,1)
      3022 CONTINUE
      DO 3023 I-1,6
      AMATA(2,1+8)=AMATA(I+8,2)
      3023 CONTINUE
      AMATO(2,JULESS+1)=SS0(JULESS-1,JJ)
      AMATA(3,JULESS+2)=SS0(JULESS-1,JJ)
      DO 3030 I-1,5
      IF(I.GE.JULESS)GO TO 3025
      AMATO(I+2,JULESS+1)=SS0(I,JULESS-1,JJ)
      AMATA(I+3,JULESS+2)=SS0(I,JULESS-1,JJ)
      GO TO 3030
      3025      AMATO(I+2,JULESS+1)=SS0(I+1,JULESS-1,JJ)
      AMATA(I+3,JULESS+2)=SS0(I+1,JULESS-1,JJ)
      3030 CONTINUE
      DO 3031 I-1,6

```

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FTN 4.6+446



```

225      3031      AMAT0(JULESS+1,I+1)-AMAT0(I+1,JULESS+1)
          CONTINUE
          DO 3032 I=1,7
            AMAT0(JULESS+2,I+2)-AMAT0(I+2,JULESS+2)
          3032      CONTINUE
          AMAT0(I+8,JULESS+7)-S1(JULESS-1,JJ)
          74.74 OPT-1
          PROGRAM CHOU
          AMAT0(I+9,JULESS+8)-S1(JULESS-1,JJ)
          DO 3040 I=1,5
            IF(I.EQ.JULESS)GO TO 3035
            AMAT0(I+8,JULESS+7)-S51(I,JULESS-1,JJ)
            AMAT0(I+9,JULESS+8)-S51(I,JULESS-1,JJ)
            GO TO 3040
          3035      AMAT0(I+8,JULESS+7)-S51(I+1,JULESS-1,JJ)
            AMAT0(I+9,JULESS+8)-S51(I+1,JULESS-1,JJ)
          3040      CONTINUE
          DO 3041 I=1,6
            AMAT0(JULESS+7,I+7)-AMAT0(I+7,JULESS+7)
          3041      CONTINUE
          DO 3042 I=1,6
            AMAT0(JULESS+8,I+8)-AMAT0(I+8,JULESS+8)
          3042      CONTINUE
          C
          C      FILL Z1, Z2, U1, U2 TO TEST BETA TERMS
          C
          3050      IN=0
          IN=0
          DO 9700 INC=1,63
            IF(X(INC,JD).EQ.1.0)GO TO 9500
            IN=IN+1
            DO 9100 J=1,6
              IF(J.EQ.JULESS)GO TO 9000
              IF(J.GT.JULESS)GO TO 9050
              U1(IN,J+1)-X(INC,J)
              GO TO 9100
            9000      Z1(IN,1)-X(INC,JULESS)
              GO TO 9100
            9050      U1(IN,J)-X(INC,J)
              GO TO 9100
            9100      CONTINUE
            GO TO 9700
          9500      IN=IN+1
            DO 9670 J=1,6
              IF(J.EQ.JULESS)GO TO 9600
              IF(J.GT.JULESS)GO TO 9650
              U2(IN,J+1)-X(INC,J)
              GO TO 9670
            9600      Z2(IN,1)-X(INC,JULESS)
              GO TO 9670
            9650      U2(IN,J)-X(INC,J)
              GO TO 9670
            9670      CONTINUE
            9700      CONTINUE
            DO 9800 INC=1,IM
              U1(INC,1)-1.0
            9800      CONTINUE
          C
          C      Z1, Z2, U1, U2 ARE FILLED
          C
          DO 9850 INC=1,IM
            U2(INC,1)-1.0
          9850      CONTINUE

```

```

285 1 9850 CONTINUE
      DO 8001 I=1,13
      DO 8001 J=1,63
      BDATA(I,J)=0.0
      BDATA(I,J)=0.0
      285 8001 CONTINUE
      74,74 OPT=1
      PROGRAM CHOU
      DO 8004 I=1,14
      DO 8004 J=1,63
      BDATA(I,J)=0.0
      285 8004 CONTINUE
      8104 CALL TRANS(Z1,NJJ,1,ZITRAN,63,1)
      CALL TRANS(Z2,NJJ,1,ZITRAN,63,1)
      CALL TRANS(U1,NJJ,6,UITRAN,63,6)
      CALL TRANS(U2,NJJ,6,UITRAN,63,6)
      IDGT=3
      CALL LINVEF(AMAT0,13,13,AMINU,IDGT,UORK,IER)
      295 C
      C
      C
      300 CALL MOVE(ZITRAN,1,NJJ,BMAT0,1,1,1,63,13,63)
      CALL MOVE(UITRAN,6,NJJ,BMAT0,2,1,6,63,13,63)
      CALL MOVE(ZITRAN,1,NJJ,BMAT0,1,NJJ+1,1,63,13,63)
      CALL MOVE(UITRAN,6,NJJ,BMAT0,8,NJJ+1,6,63,13,63)
      CALL MOVE(V1,NJJ,1,VMAT,1,1,63,1,63,1)
      CALL MOVE(V2,NJJ,1,VMAT,NJJ+1,1,63,1,63,1)
      305 C
      C
      C
      310 CALL UTRUFF(AMINU,BMAT0,13,13,63,13,13,ZINTER0,13,IER)
      CALL UTRUFF(ZINTER0,VMAT,13,63,1,13,63,FINAL0,13,IER)
      IDGT=3
      CALL LINVEF(AMATA,14,14,AMINU,IDGT,UORK,IER)
      315 C
      C
      C
      320 CALL MOVE(ZITRAN,1,NJJ,BMAT0,1,1,1,63,14,63)
      CALL MOVE(UITRAN,6,NJJ,BMAT0,3,1,6,63,14,63)
      CALL MOVE(ZITRAN,1,NJJ,BMAT0,2,NJJ+1,1,63,14,63)
      CALL MOVE(UITRAN,6,NJJ,BMAT0,9,NJJ+1,6,63,14,63)
      325 C
      C
      C
      330 CALL UTRUFF(AMINU,BMAT0,14,14,63,14,14,ZINTERA,14,IER)
      CALL UTRUFF(ZINTERA,VMAT,14,63,1,14,63,FINALA,14,IER)
      C0(I)=FINAL0(I,1)
      DO 5000 I=1,6
      D10(I,1)=FINAL0(I+1,1)
      CONTINUE
      5000 DO 5001 I=1,6
      D20(I,1)=FINAL0(I+7,1)
      CONTINUE
      5001 C1(I)=FINALA(I,1)
      C2(I)=FINALA(2,1)
      DO 5002 I=1,6
      D1(I,1)=FINALA(I+2,1)
      CONTINUE
      5002 DO 5003 I=1,6
      D2(I,1)=FINALA(I+8,1)
      CONTINUE
      5003
      ..

```

PAGE 6

12/11/78 13.11.38

FTN 4.6+446

340	C C	PROGRAM CHOU	CALL UMULFF(Z1,C1,NJJ,1.1,63.1,21C1,63,IER)	FTN 4.6+446	12/11/78 13.11.48	PAGE 7
			74/74 OPT-1			
345			CALL UMULFF(U1,D1,NJJ,6.1,63.6,UID1,63,IER)			
			CALL UMULFF(U1,D1,NJJ,1.1,63.1,21C0,63,IER)			
350			CALL UMULFF(U1,D1,NJJ,6.1,63.6,UID0,63,IER)			
			CALL ADD(Z1C1,UID1,NJJ,1,SUM1,63,1)			
			CALL SUBTR(SUM1,Z1C0,NJJ,1,SUM1,63,1)			
			CALL UMULFF(Z2,C2,NJJ,1.1,63.1,22C2,63,IER)			
355			CALL UMULFF(Z2,D2,NJJ,6.1,63.6,22C2,63,IER)			
			CALL UMULFF(Z2,C0,NJJ,1.1,63.1,22C0,63,IER)			
			CALL UMULFF(U2,D20,NJJ,6.1,63.6,22D20,63,IER)			
			CALL ADD(Z2C2,U2D2,NJJ,1,SUM2,63,1)			
360			CALL SUBTR(SUM2,Z2C0,NJJ,1,SUM2,63,1)			
			CALL SUBTR(SUM2,U2D20,NJJ,1,SUM2,63,1)			
			CALL SUBTR(V1,Z1C1,NJJ,1,SUM3,63,1)			
			CALL SUBTR(SUM3,UID1,NJJ,1,SUM3,63,1)			
365			CALL SUBTR(SUM4,U2D2,NJJ,1,SUM4,63,1)			
			CALL SUBTR(V2,Z2C0,NJJ,1,SUM5,63,1)			
			CALL SUBTR(SUM5,U2D20,NJJ,1,SUM5,63,1)			
			CALL NORM(SUM1,NJJ,50NE,63)			
			CALL NORM(SUM2,NJJ,STUO,63)			
			CALL NORM(SUM3,NJJ,STHREE,63)			
			CALL NORM(SUM4,NJJ,SFOUR,63)			
			CALL NORM(SUM5,NJJ,SFIVE,63)			
			IF(NJJ.GT.7)GO TO 5010			
			IF(NJJ.LT.6)GO TO 5015			
370	C C C		P-Q .LE. NJJ .LE. P			
			F-(50NE+SFIVE)/STHREE(NJJ-6)/(NJJ-6.)			
			WRITE(5,5005)JJ,JULESS,NJJ,F			
375			FORMAT(/,IX,"CURRENT INDICATOR=",I2,2X,"TESTED VARIABLE=",I1,2X,"P			
			POINTS IN SET 2=",I2,2X,"F VALUE=",F14.6)			
			GO TO 9998			
380	C C C		M > P			
			5010 F-(50NE+STUO)/(STHREE+SFOUR)*49.			
			WRITE(5,5005)JJ,JULESS,NJJ,F			
			GO TO 9998			
385	C C C		M < P-Q			
			5015 WRITE(5,5016)NJJ,JJ,JULESS			
			5016 FORMAT(IX,"SECOND SET CONTAINS",I1,I1,IX,"POINTS, INSUFFICIENT FOR			
			CALCULATION OF F. INDICATOR=",I2," VARIABLE=",I1)			
390			9998 CONTINUE			
			9999 STOP			
			END			
			SYMBOLIC REFERENCE MAP (R-1)			
			PROGRAM CHOU 74/74 OPT-1			
			FTN 4.6+446			
			12/11/78 13.11.48			
			PAGE 8			



```

1 0
2
3
4
5 SUBROUTINE TRANS(A,NR,NC,ATRN,NRD,NCD)
6   DIMENSION A(NRD,NCD),ATRN(NCD,NRD)
7
8   A IS MATRIX TO BE TRANSPOSED
9   NR AND NC IS NUMBER OF ROWS AND COLUMNS IN MATRIX A
10  NRD AND NCD IS NUMBER OF ROWS AND COLUMNS DIMENSIONED FOR A
11  ATRN IS THE TRANSPOSE MATRIX
12
13  DO 1 I=1,NC
14    DO 1 J=1,NR
15      ATRN(I,J)=A(J,I)
16    1 CONTINUE
17  RETURN
18  END

```

```

1 0
2
3
4
5 SUBROUTINE MOVE(XOA,IOR,IOC,XNA,INR,INC,NRD,NCD,INRD,INCD)
6   DIMENSION XOA(NRD,NCD),XNA(INRD,INCD)
7   MOVE FIL MATRIX XNA WITH SUBMATRICES XOA
8   XOA IS THE MATRIX TO BE MOVED
9   IOR AND IOC ARE NUMBER OF ROWS AND COLUMNS IN XOA
10  XNA IS THE MATRIX BEING FILLED
11  INR AND INC ARE THE ROW AND COLUMN POSITION THAT XOA(1,1)
12  STARTS IN MATRIX XNA
13  NRD AND NCD ARE THE NO. OF ROWS AND COLUMNS DIMENSIONED FOR XOA
14  INRD AND INCD ARE THE ROW AND COLUMN SIZES DIMENSIONED FOR MATRIX XNA
15
16  DO 10 I=1,IOR
17    DO 10 J=1,IOC
18      NR=INR+I-1
19      NC=INC+J-1
20      XNA(NR,NC)=XOA(I,J)
21    10 CONTINUE
22  RETURN
23  END

```



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CURRENT INDICATOR-1	TESTED VARIABLE-0	POINTS IN SET 2-12	F VALUE-	46.792030
CURRENT INDICATOR-1	TESTED VARIABLE-1	POINTS IN SET 2-12	F VALUE-	35.957190
CURRENT INDICATOR-1	TESTED VARIABLE-2	POINTS IN SET 2-12	F VALUE-	25.792171
CURRENT INDICATOR-1	TESTED VARIABLE-3	POINTS IN SET 2-12	F VALUE-	25.082949
CURRENT INDICATOR-1	TESTED VARIABLE-4	POINTS IN SET 2-12	F VALUE-	25.320975
CURRENT INDICATOR-1	TESTED VARIABLE-5	POINTS IN SET 2-12	F VALUE-	4.445489
CURRENT INDICATOR-1	TESTED VARIABLE-6	POINTS IN SET 2-12	F VALUE-	30.243100
CURRENT INDICATOR-2	TESTED VARIABLE-0	POINTS IN SET 2-10	F VALUE-	48.994204
CURRENT INDICATOR-2	TESTED VARIABLE-1	POINTS IN SET 2-10	F VALUE-	31.164791
CURRENT INDICATOR-2	TESTED VARIABLE-2	POINTS IN SET 2-10	F VALUE-	27.367543
CURRENT INDICATOR-2	TESTED VARIABLE-3	POINTS IN SET 2-10	F VALUE-	17.833968
CURRENT INDICATOR-2	TESTED VARIABLE-4	POINTS IN SET 2-10	F VALUE-	37.558254
CURRENT INDICATOR-2	TESTED VARIABLE-5	POINTS IN SET 2-10	F VALUE-	2.207488
CURRENT INDICATOR-2	TESTED VARIABLE-6	POINTS IN SET 2-10	F VALUE-	13.168126
CURRENT INDICATOR-3	TESTED VARIABLE-0	POINTS IN SET 2-12	F VALUE-	47.187560
CURRENT INDICATOR-3	TESTED VARIABLE-1	POINTS IN SET 2-12	F VALUE-	35.020414
CURRENT INDICATOR-3	TESTED VARIABLE-2	POINTS IN SET 2-12	F VALUE-	20.261425
CURRENT INDICATOR-3	TESTED VARIABLE-3	POINTS IN SET 2-12	F VALUE-	42.982841
CURRENT INDICATOR-3	TESTED VARIABLE-4	POINTS IN SET 2-12	F VALUE-	.434402
CURRENT INDICATOR-3	TESTED VARIABLE-5	POINTS IN SET 2-12	F VALUE-	1.663537
CURRENT INDICATOR-3	TESTED VARIABLE-6	POINTS IN SET 2-12	F VALUE-	2.142207
CURRENT INDICATOR-4	TESTED VARIABLE-0	POINTS IN SET 2-12	F VALUE-	25.037979
CURRENT INDICATOR-4	TESTED VARIABLE-1	POINTS IN SET 2-12	F VALUE-	4.028575
CURRENT INDICATOR-4	TESTED VARIABLE-2	POINTS IN SET 2-12	F VALUE-	43.242303
CURRENT INDICATOR-4	TESTED VARIABLE-3	POINTS IN SET 2-12	F VALUE-	42.912487
CURRENT INDICATOR-4	TESTED VARIABLE-4	POINTS IN SET 2-12	F VALUE-	35.336394
CURRENT INDICATOR-4	TESTED VARIABLE-5	POINTS IN SET 2-12	F VALUE-	12.666487
CURRENT INDICATOR-4	TESTED VARIABLE-6	POINTS IN SET 2-12	F VALUE-	.138342

SECOND SET CONTAINS 4 POINTS. INSUFFICIENT FOR CALCULATION OF F. INDICATOR-5 VARIABLE-0  
SECOND SET CONTAINS 4 POINTS. INSUFFICIENT FOR CALCULATION OF F. INDICATOR-5 VARIABLE-1

00  
00



SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-5	VARIABLE-2
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-5	VARIABLE-3
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-5	VARIABLE-4
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-5	VARIABLE-5
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-5	VARIABLE-6
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-6	VARIABLE-0
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-6	VARIABLE-1
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-6	VARIABLE-2
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-6	VARIABLE-3
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-6	VARIABLE-4
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-6	VARIABLE-5
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-6	VARIABLE-6
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-7	VARIABLE-0
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-7	VARIABLE-1
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-7	VARIABLE-2
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-7	VARIABLE-3
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-7	VARIABLE-4
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-7	VARIABLE-5
SECOND SET CONTAINS 4 POINTS.	INSUFFICIENT FORCALCULATION OF F.	INDICATOR-7	VARIABLE-6
CURRENT INDICATOR-8	TESTED VARIABLE-0 POINTS IN SET 2-8	F VALUE-	49.001918
CURRENT INDICATOR-8	TESTED VARIABLE-1 POINTS IN SET 2-8	F VALUE-	36.232245
CURRENT INDICATOR-8	TESTED VARIABLE-2 POINTS IN SET 2-8	F VALUE-	36.429398
CURRENT INDICATOR-8	TESTED VARIABLE-3 POINTS IN SET 2-8	F VALUE-	47.933922
CURRENT INDICATOR-8	TESTED VARIABLE-4 POINTS IN SET 2-8	F VALUE-	44.038580
CURRENT INDICATOR-8	TESTED VARIABLE-5 POINTS IN SET 2-8	F VALUE-	48.997518
CURRENT INDICATOR-8	TESTED VARIABLE-6 POINTS IN SET 2-8	F VALUE-	35.318639
CURRENT INDICATOR-9	TESTED VARIABLE-0 POINTS IN SET 2-51	F VALUE-	49.440065
CURRENT INDICATOR-9	TESTED VARIABLE-1 POINTS IN SET 2-51	F VALUE-	40.300204
CURRENT INDICATOR-9	TESTED VARIABLE-2 POINTS IN SET 2-51	F VALUE-	14.671381
CURRENT INDICATOR-9	TESTED VARIABLE-3 POINTS IN SET 2-51	F VALUE-	45.948176
CURRENT INDICATOR-9	TESTED VARIABLE-4 POINTS IN SET 2-51	F VALUE-	.735032
CURRENT INDICATOR-9	TESTED VARIABLE-5 POINTS IN SET 2-51	F VALUE-	7.367655
CURRENT INDICATOR-9	TESTED VARIABLE-6 POINTS IN SET 2-51	F VALUE-	.680680
CURRENT INDICATOR-10	TESTED VARIABLE-0 POINTS IN SET 2-26	F VALUE-	48.900573
CURRENT INDICATOR-10	TESTED VARIABLE-1 POINTS IN SET 2-26	F VALUE-	35.818197
CURRENT INDICATOR-10	TESTED VARIABLE-2 POINTS IN SET 2-26	F VALUE-	43.051565
CURRENT INDICATOR-10	TESTED VARIABLE-3 POINTS IN SET 2-26	F VALUE-	21.841772
CURRENT INDICATOR-10	TESTED VARIABLE-4 POINTS IN SET 2-26	F VALUE-	28.714985
CURRENT INDICATOR-10	TESTED VARIABLE-5 POINTS IN SET 2-26	F VALUE-	7.509505

AD-A064 658

AIR FORCE INST OF TECH WRIGHT-PATTERSON AFB OHIO SCH--ETC F/G 12/1  
CRITERION FOR SELECTION OF VARIABLES IN A REGRESSION ANALYSIS.(U)

DEC 78 L J PULCHER

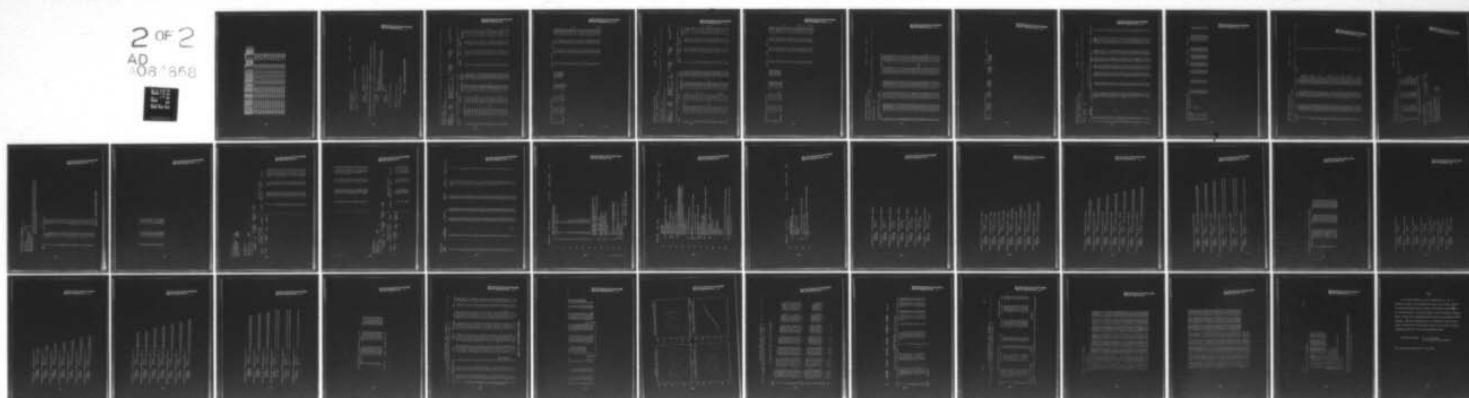
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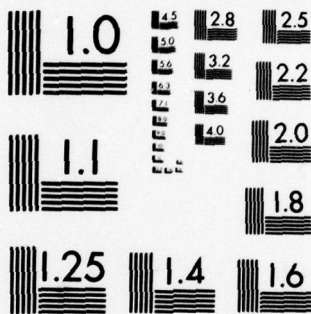
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MICROCOPY RESOLUTION TEST CHART  
NATIONAL BUREAU OF STANDARDS-1963-A



CURRENT INDICATOR-10	TESTED VARIABLE-6	POINTS IN SET 2-26	F VALUE-	19.336805	VARIABLE- 0
SECOND SET CONTAINS 4	POINTS. INSUFFICIENT FORCALCULATION OF F.	INDICATOR- 11	INDICATOR- 11	VARIABLE- 1	
SECOND SET CONTAINS 4	POINTS. INSUFFICIENT FORCALCULATION OF F.	INDICATOR- 11	INDICATOR- 11	VARIABLE- 2	
SECOND SET CONTAINS 4	POINTS. INSUFFICIENT FORCALCULATION OF F.	INDICATOR- 11	INDICATOR- 11	VARIABLE- 3	
SECOND SET CONTAINS 4	POINTS. INSUFFICIENT FORCALCULATION OF F.	INDICATOR- 11	INDICATOR- 11	VARIABLE- 4	
SECOND SET CONTAINS 4	POINTS. INSUFFICIENT FORCALCULATION OF F.	INDICATOR- 11	INDICATOR- 11	VARIABLE- 5	
SECOND SET CONTAINS 4	POINTS. INSUFFICIENT FORCALCULATION OF F.	INDICATOR- 11	INDICATOR- 11	VARIABLE- 6	
CURRENT INDICATOR-12	TESTED VARIABLE-0	POINTS IN SET 2-23	F VALUE-	54.983835	
CURRENT INDICATOR-12	TESTED VARIABLE-1	POINTS IN SET 2-23	F VALUE-	42.912617	
CURRENT INDICATOR-12	TESTED VARIABLE-2	POINTS IN SET 2-23	F VALUE-	45.636063	
CURRENT INDICATOR-12	TESTED VARIABLE-3	POINTS IN SET 2-23	F VALUE-	35.171661	
CURRENT INDICATOR-12	TESTED VARIABLE-4	POINTS IN SET 2-23	F VALUE-	15.781559	
CURRENT INDICATOR-12	TESTED VARIABLE-5	POINTS IN SET 2-23	F VALUE-	17.990961	
CURRENT INDICATOR-12	TESTED VARIABLE-6	POINTS IN SET 2-23	F VALUE-	15.147350	
CURRENT INDICATOR-13	TESTED VARIABLE-0	POINTS IN SET 2-20	F VALUE-	46.647943	
CURRENT INDICATOR-13	TESTED VARIABLE-1	POINTS IN SET 2-20	F VALUE-	39.790991	
CURRENT INDICATOR-13	TESTED VARIABLE-2	POINTS IN SET 2-20	F VALUE-	22.917883	
CURRENT INDICATOR-13	TESTED VARIABLE-3	POINTS IN SET 2-20	F VALUE-	39.312276	
CURRENT INDICATOR-13	TESTED VARIABLE-4	POINTS IN SET 2-20	F VALUE-	11.509766	
CURRENT INDICATOR-13	TESTED VARIABLE-5	POINTS IN SET 2-20	F VALUE-	15.344824	
CURRENT INDICATOR-13	TESTED VARIABLE-6	POINTS IN SET 2-20	F VALUE-	11.583250	

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11/26/79 12.35.34. PAGE 1

VOSEBACH COMPUTING CENTER  
NORTHWESTERN UNIVERSITY

S P S S - STATISTICAL PACKAGE FOR THE SOCIAL SCIENCES

VERSION 7.0 -- JUNE 27 1977

QMC NAME  
DATA NAME  
CONTROL  
VARIABLE LIST V1 TO V93  
INPUT FILE FIVE02(12X,3F20.14),124,2F23.14)

THE INPUT FORMAT PROVIDES FOR 94 VARIABLES. 91 WILL BE READ  
IT PROVIDED FOR 13 RECORDS (\*CARDS\*) PER CASE. A MAXIMUM OF 72 \*COLUMNS\* ARE USED ON A RECORD.

WARNING - A NUMERIC VARIABLE HAS A WIDTH GREATER THAN 14. SMALL ROUNDING/TRUNCATION ERRORS MAY OCCUR.

TYPE CASES  
INPUT METHOD  
REGRESSION  
METHODS: STEPWISE/  
VARIABLES: V1 TO V42, V52 TO V62/  
REGRESSION: V43 WITH V1 TO V42, V52 TO V62, V72 TO V93,  
V73 TO V77, V52 TO V73, V74 TO V77, V78, V56 TO V73, V93/  
RESULTS

OPTIONS  
STATISTICS  
READ INPUT DATA  
CRITERIA

OPTION 100 ON NEED FOR REGRESSION

OPTION - 1  
10000 MISSING VALUE INDICATORS

OPTION - 15  
PRINT UNNORMALIZED REGRESSION COEFFICIENTS

Figure 7. Selected Output from SPSS

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MULTIPLE REGRESSION									
DEPENDENT VARIABLE.. V03									
VARIABLE(S) ENTERED ON STEP NUMBER 24.. V23									
MULTIPLE R									
R SQUARE									
ADJUSTED R SQUARE									
STD DEVIATION									
ANALYSIS OF VARIANCE									
RESIDUAL									
COEFF OF VARIABILITY									
SUM OF SQUARES									
MEAN SQUARE									
F									
SIGNIFICANCE									
VARIABLES IN THE EQUATION									
SIGNIFICANCE									
VARIABLES NOT IN THE EQUATION									
SIGNIFICANCE									
VARIABLE	B	STD ERROR B	F	ELASTICITY	BTB	VARIABLE	PARTIAL TOLERANCE	F	SIGNIFICANCE
V5	.42322761E-01	.32973206	.1736357E-01	.0730174	.0730174	V2	.04743	.05293	.2004374E
V28	-.21210408	.12234727	3.0993134	-.15514	-.15514	V4	-.24762	.26009	2.54742E-1
V11	-1.4539365	.3503280	17.22414	-.21574	-.21574	V6	.17585	.29350	1.24454E-1
V64	3.115064	.61655375	12.072175	2.205243	2.205243	V7	.41348	.04077	9.15556E-0
V8	19.184504	2.6540934	17.741777	2.312591	2.312591	V9	.04227	.33436	1.022255E-01
V23	.27771112	.55941272E-01	10.373727	-.124354	-.124354	V12	-.32395	.50216	.1119401
VK3	-2.0777036	.5783844	12.412051	-.155232	-.155232	V13	-.07450	.90750	.2417472E
V35	-.52681783	.11530701	20.502056	4.25119	4.25119	V15	-.11107	.01611	.6925870E
V27	-1.7076150	.7243465	6.714334	1.73404	1.73404	V18	.26115	.01552	2.65361E-1
V15	-1.1362296	1.1353329	7.711510	-.954163	-.954163	V19	.02356	.06440	.2221674E-01
V17	-.7735907E-01	.30023068	.3076731E-01	3.23105	3.23105	V20	.50195	.04926	22.10237E
V34	.5444439	.10701779	12.329178	.553102	.553102	V21	.57763	.04456	15.35533E
V10	-1.2884946	.32202795	15.505080	-.712522	-.712522	V22	.35113	.06733	5.4306740E
V1	.5103312	.13720647	13.077111	.34902	.34902	V25	.32427	.06430	11.79001E
V5	.7161154	.76420443E-01	15.532361	-.113765	-.113765	V29	-.35820	.02050	.1355517E
V16	-.37240501	.10336377	3.1125623	-.275730	-.275730	V32	.01949	.30042	.4291320E-02
V28	2.7715473	.85424161	7.193522	1.775511	1.775511	V33	.07282	.31057	.2301211E-01
V25	.62155458	.17697139	5.274727	-.25502	-.25502	V34	.01619	.31358	.1022070E-01
			.071	-.3.05143	-.3.05143				.320



SPCSSTREPOISE	11/24/78	12.35.34.	PAGE	63
V25	-6.7951336	.120.5924	.06249	.00518
V75	5.14-9273	.79731950	.15125	.02457
V15	3.024-1170	1.7035243	.21201	.04453
V23	-6.1276-355-01	.723016105-01	.21737	.00975
(CONSTANT)	-5.119146	.73713092	.25557	.00452
			.17711	.00534
			.16317	.00001
			.04515	.01074
			.03074	.01475
			.12013	.01003
			.15037	.01667
			.11566	.02015
			.08102	.02557
			.33132	.01953
			.22544	.02115
			.21201	.02531
			.11375	.03776
			.02931	.02370
			.02967	.02107

11/24/78	12.35.34.	PAGE	63
V73	.06249	.00518	.26900220
V43	.15125	.02457	.01307010
V41	.21201	.04453	1.8499076
V42	.21737	.00975	1.9341358
V62	.25557	.00452	2.7322170
V63	.17711	.00534	1.2633375
V66	.16317	.00001	1.0563216
V67	.04515	.01074	1.0730544-31
V69	.03074	.01475	1.1371752-71
V74	.12013	.01003	.1729414
V76	.15037	.01667	1.5042515
V77	.11566	.02015	1.5287765
V79	.08102	.02557	1.4577120
V85	.33132	.01953	4.5071115
V87	.22544	.02115	2.0961274
V89	.21201	.02531	2.1315661
V91	.11375	.03776	1.1711115
V92	.02931	.02370	1.2506927-71
V93	.02967	.02107	1.3435129-71

SPSSSTEPWISE WITH 60 LEFT FROM CHOW

FILE NAME (CREATION DATE = 11/24/73)

DEPENDENT VARIABLE.. V03

VARIABLE(S) ENTERED ON STEP NUMBER 25.. V20

MULTIPLE R .97577  
R SQUARE .95212  
ADJUSTED R SQUARE .92366  
STD DEVIATION .47827

ANALYSIS OF VARIANCE  
REGRESSION 22.  
RESIDUAL 35.  
COEFF OF VARIABILITY 116.5 PDI

SUM OF SQUARES  
140.96044  
7.49108

MEAN SQUARE  
5.4754  
.19268

F 33.71811  
SIGNIFICANCE

..... MULTIPLE REGRESSION .....

VARIABLES IN THE EQUATION				VARIABLES NOT IN THE EQUATION			
VARIABLE	B	STD ERROR B	F SIGNIFICANCE	VARIABLE	PARTIAL TOLERANCE	F SIGNIFICANCE	
V3	.85484705-01	.263567	.10290231	V2	.03225	.05295	.24572645
V56	-.27214195	.59760343E-01	.783	V4	.13050	.12998	.023
V11	-1.4370595	.22310639	.210	V6	.09241	.24445	.1357135
V64	3.6134969	.72524203	.030	V7	-.21251	.01151	.27031645
V8	11.725694	2.3205911	.000	V9	.21312	.31112	.305
V89	.79487513	.59154103E-01	.000	V12	.03649	.57439	1.6701220
V63	-1.0056666	.4672035	.003	V13	-.13940	.09530	.187
V65	-.6427356	.57022032E-01	.030	V15	-.01519	.01553	.3708234
V27	-1.5113210	.58014751	.004	V19	-.07292	.01137	.7130163
V16	-2.2505732	.56079748	.004	V19	-.01524	.00615	.87759307E-02
V17	-.22715744E-01	.31612660	.004	V21	.02667	.01037	.2651127
V94	.59769672	.17712608	.071	V22	.03917	.04863	.10326162E-01
V10	-1.1355451	.25399375	.030	V26	.13751	.02016	.27345775E-01
V1	.30278193	.10907972	.001	V29	-.05190	.01999	.50327125E-01
V5	.51240037	.67542045E-01	.001	V32	.10742	.26523	.73241055
V90	-.44515539	.91453075E-01	.000	V31	.15004	.20613	.10201791
V29	7.5068232	.71535773	.000	V34	.15741	.30747	.13442369
V75	.42277740	.14312100	.005	V39	.13995	.00930	.28704432
							.11490325

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	11/24/75	12.35.34.	240F	70
V25	-1.302062	.16222	.0254	1.0122008
V26	2.31-25	.17112	.04355	.315
V27	.753770	.20654	.09638	1.152221
V28	-2.65563	.10211	.00659	.299
V29	.783242	-.15012	.00529	1.633024
V30	-1.35201	-.09117	.00801	.201
V31	-.763719	.01413	.01071	1.025535
V32	1.55.47	.01527	.01175	.313
V33	.7139581	-.05114	.00331	.67445769
V34	-1.43233	-.05212	.01556	.22135102
V35		.11310	.01791	.22135102
V36		.07579	.02513	.574
V37		.00919	.01377	.62579015-12
V38		-.09018	.01623	.11343674
V39		-.033910	.01940	.735
V40		-.06058	.03205	.14401103
V41		-.02765	.02354	.706
V42		-.06304	.02101	.10349124
V43				.57
V44				.923363
V45				.407
V46				.482014425-11
V47				.27
V48				.342290015-12
V49				.554
V50				.7115.340
V51				.540
V52				.19371525-11
V53				.203
V54				.2835.354
V55				.297
V56				.273260965-11
V57				.064
V58				.50238575-13
V59				.932



SSSSSTP015015 WITH 60 LEFT FROM CHOW

11/24/78 12.35.34. PAGE 143

FILE NAME (CREATION DATE = 11/24/78)

DEPENDENT VARIABLE.. V94

..... MULTIPLE REGRESSION .....

COEFFICIENTS AND CONFIDENCE INTERVALS.

VARIABLE	B	STD. ERROR B	T	95.0 PCI CONFIDENCE INTERVAL
V7	-2.2566830	1.3557844	-1.6596376	-5.1730446
V8	-1.195714	4.5932539	-3.6795910	-1.1267297
V11	-1.5122731	2.564724	-2.7644022	-2.3207193
V12	3.2522220	1.7556818	2.0550174	1.1155252
V6	11.342682	6.251157	3.6533000	2.4570177
V9	1.6320539	2.754253	2.579720	2.2565031
V23	-1.6322134	2.2718560	-1.626327	-2.5169241
V61	-2.2760312	3.634726	-7.122662	-2.5113374
V27	-2.0966039	1.0217014	-2.0463724	-5.2749361
V17	-7.3026031	1.1102453	-2.9563729	-5.6377391
V1	1.355925	2.281480	1.740111	1.223393
V3	4.131217	1.147053	1.244143	1.6457253
V90	-2.334482	2.2539610	-1.0332153	1.0311171
V29	5.136381	1.7612230	2.8832756	2.3333643
V72	-7.255131	4.450621	-1.6476613	-1.7723212
V25	-1.476129	4.195281	-2.7203955	-2.6767652
V70	1.5267724	1.1944260	1.2717330	1.2379733
V14	-2.180275	2.4378656	-1.2630944	-2.3529643
V23	-5.116077	2.2211204	-3.4661044	-1.3317351
V20	-5.116079	4.027643	-1.5731364	-2.601303
V53	-2.126136	5.625876	-1.522273	-2.4034802
V23	1.1776106	3.655920	3.1954779	2.7116792
V52	-5.750223	4.237723	-1.3701639	-2.1712561
V50	3.756217	1.214010	2.5720940	2.2928771
V41	-9.447611	2.712036	-1.7739237	-2.2047403
V7	-4.344646	1.125942	-3.9631124	-1.5259380
V92	1.859157	3.0119548	3.1727254	1.6903033
V31	-1.037483	3.762823	-2.721395	-1.511155
V6	5.227835	2.567128	2.022684	1.5214051
V77	1.271977	2.532136	2.179032	1.1716913
V2	1.171642	2.721326	1.515661	1.552137
V13	-1.197813	1.7205786	-4.212870	-2.2727371
V73	-1.128787	2.785533	-1.513169	1.1613796
V12	-5.240478	4.4842463	-1.1729175	-1.3616525
V74	-6.571577	3.137875	-2.1128235	-1.3446113
V77	3.204430	2.1174220	1.364953	1.2143993
V71	1.2963437	2.2469799	1.0428176	1.3713022
V55	-5.673972	3.721077	-1.577032	-2.5710991
V18	5.205452	3.177182	1.638273	2.4567283
V15	2.776249	2.7397226	1.0131241	1.672701
V4	3.770817	3.141724	1.199669	1.615600
V22	-3.324156	3.141726	-3.123722	-2.7770154
V75	3.552744	6.937171	0.514120	1.123053
V29	2.149620	6.645869	0.321351	1.5112559
V19	1.254522	3.710779	0.332553	1.7507415
V2	-1.111636	3.691239	-0.764684	-1.7319073

-01  
-02  
-01

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SESSIES	11/24/79	12.35.14.	PAGE 144
VT	-3.510.43		
VT	2.558.134		
CO	-1.130.932		
CO	1.431.177		
CO	-1.632.739		
CO	2.019.926		

SOSSTERWISE WITH 56 LEFT FROM CHOW  
FILE NAME (CREATION DATE = 11/24/73)  
..... MULTIPLE REGRESSION .....  
DEPENDENT VARIABLE.. V93

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SUMMARY TABLE

STCO	VARIABLE ENTERED	ENTER OR REMOVE	SIGNIFICANCE	MULTIPLE R	R SQUARE CHANGE	SIMPLE R	OVERALL F	SIGNIFICANCE
1	V3	81.00022	0	.75134	.57312	.75704	41.83622	0
2	V4	15.00000	.000	.81131	.69870	.77304	17.91019	.000
3	V11	13.21833	.001	.84372	.72117	-.35276	50.86727	.000
4	V6	9.53350	.003	.87203	.78004	.27332	46.05272	.000
5	V4	6.55326	.014	.88378	.78460	.60361	41.52843	.000
6	V43	5.74072	.020	.89310	.80354	.4957	31.17560	.000
7	V43	4.77632	.041	.90364	.81602	.26382	29.31799	.000
8	V93	3.44330	.067	.91342	.82905	-.01576	32.77534	.000
9	V28	3.01940	.084	.91819	.83450	.11209	30.57272	.000
10	V15	3.73372	.085	.92183	.84031	.24104	28.71499	.000
11	V15	4.71552	.035	.92558	.84607	-.24230	25.91659	.000
12	V17	4.15334	.047	.9317	.87067	.13040	23.5321	.000
13	V3	1.15005	.217	.93287	.87801	.13374	26.77726	.000
14	V11	1.33132	.200	.93312	.88000	.1357	24.13724	.000
15	V1	1.33267	.165	.93494	.88494	.44014	24.0319	.000
16	V1	1.27545	.268	.94213	.89799	.22783	22.77793	.000
17	V93	2.63974	.096	.94301	.8975	.13576	22.5025	.000
18	V23	2.33253	.124	.94341	.90003	.13570	22.5025	.000
19	V15	1.0315	.971	.94371	.90065	.23050	23.6472	.000
20	V75	2.3201	.154	.94375	.90067	-.14500	23.6472	.000
21	V25	1.63027	.204	.94375	.90067	.03110	22.4320	.000
22	V73	2.21212	.144	.94375	.90067	.23050	22.4320	.000
23	V17	2.44112	.123	.94375	.90067	.23050	22.4320	.000
24	V23	2.60332	.085	.94372	.90067	.23050	22.4320	.000
25	V3	2.13212	.000	.94372	.90067	.23050	22.4320	.000
26	V7	2.13212	.197	.94372	.90067	.23050	22.4320	.000
27	V7	2.13212	.152	.94372	.90067	.23050	22.4320	.000
28	V3	1.63027	.101	.94372	.90067	.23050	22.4320	.000
29	V23	1.33113	.174	.94372	.90067	.23050	22.4320	.000
30	V2	1.63027	.205	.94372	.90067	.23050	22.4320	.000
31	V19	1.35943	.212	.94372	.90067	.23050	22.4320	.000
32	V1	5.11016	.031	.94372	.90067	.23050	22.4320	.000
33	V7	2.63217	.104	.94372	.90067	.23050	22.4320	.000
34	V42	1.25293	.272	.94372	.90067	.23050	22.4320	.000
35	V91	1.3307	.427	.94372	.90067	.23050	22.4320	.000
36	V7	1.13752	.295	.94372	.90067	.23050	22.4320	.000
37	V2	1.15242	.290	.94372	.90067	.23050	22.4320	.000
38	V13	.74341	.381	.94372	.90067	.23050	22.4320	.000
39	V73	.61634	.581	.94372	.90067	.23050	22.4320	.000
40	V73	.73030	.363	.94372	.90067	.23050	22.4320	.000
41	V12	.63042	.884	.94372	.90067	.23050	22.4320	.000
42	V7	.63042	.421	.94372	.90067	.23050	22.4320	.000
43	V74	.63042	.433	.94372	.90067	.23050	22.4320	.000
44	V33	1.11572	.900	.94372	.90067	.23050	22.4320	.000
45	V75	.52401	.207	.94372	.90067	.23050	22.4320	.000
46	V75	.52401	.474	.94372	.90067	.23050	22.4320	.000



SPSSSTEPWISE WITH 60 LEFT FROM CHOW

FILE 'D:\NAME (CREATION DATE = 11/24/78)

LINE	NAME	VALUE
47	V55	.45336
48	V14	.1121
49	V15	.67726
50	V16	.5004
51	V17	.9211
52	V18	.0755
53	V19	.5863
54	V20	.0117
55	V21	.2441
56	V22	.1267
57	V23	.6511
58	V24	.0139
59	V25	.0377

LINE	NAME	11/24/78	12.15.78	PAGE	149
47	V55	.52062	.00013	.53196	25.91009
48	V14	.92101	.00039	.56152	24.59259
49	V15	.95165	.03665	.24005	23.55161
50	V16	.90166	.00000	.13127	26.65459
51	V17	.43255	.00015	.52379	24.53323
52	V18	.90215	.00000	.22220	26.81725
53	V19	.92307	.00022	.69379	25.65173
54	V20	.95161	.00071	.09370	24.25420
55	V21	.93161	.00500	.15538	26.51272
56	V22	.58303	.00022	.07160	24.89670
57	V23	.98495	.00012	.17214	23.10251
58	V24	.90401	.00003	.26072	21.40637
59	V25	.92303	.00003	.26715	13.74265
60	V26	.93413	.00004	.15034	12.39196

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SPSSSTEPWISE 4114 50 LEFT FROM CHOW

11/24/78 12.35.74. PAGE 150

FILE NAME (CREATION DATE = 11/24/78)

MULTIPLE REGRESSION				RESIDUAL	
OBSERVATION	Y VALUE	Y ESTIMATE	RESIDUAL	-2SD	+2SD
1.	-521.149	-7621.779	.2706-01		
2.	-765.724	-11671.35	-.245937		
3.	-1732.51	-15327.51	-.2024065-01		
4.	1.80392	1.65594	-.566463E-01		
5.	159.416	12743.94E-01	.817472E-01		
6.	-386.674	-6037.63	-.103103E-01		
7.	-2.96572	-2.54391	-.165114E-01		
8.	-3.23717	-3.19673	-.64243E-01		
9.	-554.624	-776.227	.117114E		
10.	-745.139	-1173.72	-.11175E-01		
11.	-2.66704	-2.55734	.109411E-01		
12.	597.722	6013.91	-.51193E-02		
13.	-1.74525	-1.91533	.162037		
14.	1.26321	1.33784	-.643174E-01		
15.	1.10132	1.26287	-.156564		
16.	-1.476109	-1.34715	.1651159		
17.	-2.35220	-2.29642	-.5270327E-01		
18.	-1.01535	-1.02812	-.151751E-01		
19.	2.165391	7.857363	.1476121		
20.	2.678142	2.79745	-.114735E-01		
21.	-1.69102	-1.92521	-.246437		
22.	2.97271	1.74217	.165713		
23.	1.27149	1.26075	.6224-0E-01		
24.	1.13342	1.62932	.105157		
25.	-7.16135	-9.22763E-02	-.922763E-02		
26.	3.26234	3.25112	.231796E-01		
27.	-1.06743	2.11182	-.316439		
28.	1.13473	1.17152	-.330217E-01		
29.	1.36132	1.74739	.268222E-01		
30.	-2.02293	-3.27263	.120651		
31.	-1.69116	-1.56813	.77293E-01		
32.	928.774	1.21765	-.232259		
33.	-3767.40	-1.07620	.637763E-01		
34.	-2.565128	-2.64307	.127672		
35.	2.16727	-2.61780	.540131E-01		
36.	2013.19	1274.20	.771013E-01		
37.	-2064.82	-2.91194	.267033E-01		
38.	-1.66403	-1.55145	.35471E-01		
39.	-2.26513	-2.159534	-.120544		
40.	817.307	7.25249	.328532E-01		
41.	-321.554	-2.99473	-.25677E-01		
42.	1.36123	1.31624	.272236E-01		
43.	-1.87431	.131495	.936113E-02		
44.	-364.770	-.811224	-.721516E-01		
45.	2.64358	.273203	-.110391E-01		
46.	-5.66666	-.652374	-.467624		
47.	-2217.56	-.576767	.110771		
48.	-2761.65	-.652173	.326147		
49.	-2767.43	-2.760574	-.563243E-01		
50.	-1.551405	-1.630365	.360793E-01		
51.	641.121	.193313	.57273E-01		
52.	133.101	.113193	-.163974		

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SPRINTERLINE WITH 60 LEFT FROM CYON

FILE NAME (CREATION DATE = 11/24/78)

CONSERVATION	Y VALUE	Y ESTIMATE	RESIDUAL	-250	0.0	+250
51.	.586160	.476320	.109840		1	
52.	.585270	.476320	.108950		1	
53.	1.235770	1.027117	.208653		1	
54.	1.011500	1.027117	-.015617		1	
55.	.734110	1.027117	-.293007		1	
56.	.166570	1.027117	-.860547		1	
57.	.102110	1.027117	-.925007		1	
58.	.811110	1.027117	-.216007		1	
59.	.811110	1.027117	-.216007		1	
60.	.209200	1.027117	-.817917		1	
61.	.471110	1.027117	-.556007		1	
62.	.103670	1.027117	-.923447		1	
63.	.103670	1.027117	-.923447		1	

NOTE - (\*) INDICATES ESTIMATE CALCULATED WITH MEANS SUBSTITUTED  
 2 INDICATES POINT OUT OF RANGE OF DATA

NUMBER OF CASES PLOTTED	63.	1. OR 1.54 PERCENT OF THE TOTAL
NUMBER OF S.D. OUTLIERS	1. OR 1.54 PERCENT OF THE TOTAL	
VON NEUMANN RATIO	2.23711	DURBIN-WATSON TEST 2.25030
NUMBER OF POSITIVE RESIDUALS	33.	
NUMBER OF NEGATIVE RESIDUALS	30.	
NUMBER OF ZERO RESIDUALS	35.	
EXPECTED NUMBER OF SIGNS	33.	
EXCEEDED S.D. OF SIGN DISTRIBUTION	3.32702	
UNIT NORMAL DEVIATE	.79205	
PROBABILITY OF OBTAINING .65. POS(2)	.21709	

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2.0121  
3.5175  
2.3157  
1.5109  
4.5273  
1.3232  
1.5237  
2.3111  
2.7103  
4.4215  
3.5222  
1.2183  
3.2130  
1.9132  
2.7125  
4.5109  
3.1350  
1.5332  
2.0321  
1.3152

1.127  
2.0121  
3.5175  
2.3157  
1.5109  
4.5273  
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3.2130  
1.9132  
2.7125  
4.5109  
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SUB-ROUTINE 1  
DEPENDENT VARIABLE 61  
MAXIMUM NUMBER OF STEPS 122  
F-LEVEL FOR INCLUSION .110010  
F-LEVEL FOR DELETION .000000  
TOLERANCE LEVEL .001000

STEP NUMBER 1  
VARIABLE ENTERED 1  
MULTIPLE R .7670  
STD. ERROR OF EST. 1.0464

ANALYSIS OF VARIANCE  
REGRESSION 1  
RESIDUAL 61  
OF SUM OF SQUARES  
89.555  
60.797  
MEAN SQUARE  
89.555  
1.043  
F RATIO  
81.596

VARIABLES IN EQUATION			VARIABLES NOT IN EQUATION		
VARIABLE	COEFFICIENT	STD. ERROR	F TO REMOVE	PARTIAL CORR.	TOLERANCE
CONSTANT	-1.37339721				
1	1.11106	.12351	81.8362 (2)	.10129	.637
2				.02626	.112
3				.0317	.041
4				.02227	.0905
5				.02672	.043
6				.0337	.0301
7				.0339	.0339
8				.0339	.0339
9				.0339	.0339
10				.0339	.0339
11				.0339	.0339
12				.0339	.0339
13				.0339	.0339
14				.0339	.0339
15				.0339	.0339
16				.0339	.0339
17				.0339	.0339
18				.0339	.0339
19				.0339	.0339
20				.0339	.0339
21				.0339	.0339
22				.0339	.0339
23				.0339	.0339
24				.0339	.0339
25				.0339	.0339
26				.0339	.0339
27				.0339	.0339
28				.0339	.0339
29				.0339	.0339
30				.0339	.0339
31				.0339	.0339
32				.0339	.0339
33				.0339	.0339



11.41	5.50	6.03	120
-2.11	3.50	3.44	120
-1.22	3.50	2.56	120
-2.31	3.50	3.51	120
-2.70	3.50	4.76	120
-2.83	3.50	4.51	120
-2.90	3.50	4.31	120
-2.55	3.50	4.32	120
-3.10	3.50	11.03	120
-3.31	3.50	11.33	120
-3.52	3.50	15.77	120
-3.52	3.50	1.72	120
-3.52	3.50	2.35	120
-3.52	3.50	1.31	120
-3.52	3.50	1.11	120
-3.52	3.50	2.10	120
-3.52	3.50	3.25	120
-3.52	3.50	7.59	120
-3.52	3.50	3.13	120
-3.52	3.50	2.56	120
-3.52	3.50	2.72	120
-3.52	3.50	3.38	120
-3.52	3.50	3.00	120
-3.52	3.50	3.75	120
-3.52	3.50	3.25	120
-3.52	3.50	3.21	120

STEP NUMBER 2  
USAIN 578125A  
VARIABLE ENTERED 17

MULTIPLE R  
SYD. ERROR OF EST.  
.9434  
.9116

ANALYSIS OF VARIANCE

OF	SUM OF SQUARES	MEAN SQUARE
REGRESSION	2	108.65
RESIDUAL	60	53.396
		51.595
		.593

# VALUES IN EQUATION

CYC-FICIFNY STT. E2222 = TO REMOVE

CONSTANT  
3  
43

-4.62472477		
1.07143	.11211	91.3258 (2)
.12758	.03215	15.0462 (2)

1.011 0003 61 100. 53761876A

PARTIAL COR.

3 1 1 121

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SUMMARY TABLE

STEP NUMBER	VARIABLE ENTERED REMOVED	R	MULTIPLE	INCREASE IN KSO	F VALUE TO ENTER OR REMOVE	HUNDREDS OF INDEPENDENT VARIABLES INCLUDED
1	3	.7870	.5731	.5731	81.332	1
2	41	.9116	.7337	.0000	11.042	2
3	11	.8132	.7012	.0025	17.243	3
4	30	.9721	.7505	.0000	2.772	4
5	0	.8813	.7115	.0000	2.352	5
6	54	.9074	.7335	.0000	2.839	6
7	38	.9015	.7180	.0000	2.871	7
8	60	.9125	.7211	.0000	3.003	8
9	26	.9177	.7185	.0000	3.003	9
10	15	.9216	.7132	.0000	3.003	10
11	16	.9205	.7132	.0000	3.003	11
12	17	.9205	.7132	.0000	3.003	12
13	17	.9205	.7132	.0000	3.003	13
14	10	.9211	.7132	.0000	3.003	14
15	1	.9211	.7132	.0000	3.003	15
16	1	.9211	.7132	.0000	3.003	16
17	1	.9211	.7132	.0000	3.003	17
18	1	.9211	.7132	.0000	3.003	18
19	1	.9211	.7132	.0000	3.003	19
20	1	.9211	.7132	.0000	3.003	20
21	1	.9211	.7132	.0000	3.003	21
22	1	.9211	.7132	.0000	3.003	22
23	1	.9211	.7132	.0000	3.003	23
24	1	.9211	.7132	.0000	3.003	24
25	1	.9211	.7132	.0000	3.003	25
26	1	.9211	.7132	.0000	3.003	26
27	1	.9211	.7132	.0000	3.003	27
28	1	.9211	.7132	.0000	3.003	28
29	1	.9211	.7132	.0000	3.003	29
30	1	.9211	.7132	.0000	3.003	30
31	1	.9211	.7132	.0000	3.003	31
32	1	.9211	.7132	.0000	3.003	32
33	1	.9211	.7132	.0000	3.003	33
34	1	.9211	.7132	.0000	3.003	34
35	1	.9211	.7132	.0000	3.003	35
36	1	.9211	.7132	.0000	3.003	36
37	1	.9211	.7132	.0000	3.003	37
38	1	.9211	.7132	.0000	3.003	38
39	1	.9211	.7132	.0000	3.003	39
40	1	.9211	.7132	.0000	3.003	40
41	1	.9211	.7132	.0000	3.003	41
42	1	.9211	.7132	.0000	3.003	42
43	1	.9211	.7132	.0000	3.003	43
44	1	.9211	.7132	.0000	3.003	44
45	1	.9211	.7132	.0000	3.003	45
46	1	.9211	.7132	.0000	3.003	46
47	1	.9211	.7132	.0000	3.003	47
48	1	.9211	.7132	.0000	3.003	48
49	1	.9211	.7132	.0000	3.003	49
50	1	.9211	.7132	.0000	3.003	50
51	1	.9211	.7132	.0000	3.003	51

```

1      PROGRAM MULTI(INPUT,OUTPUT,TAPE1,TAPE3)
2      GROUPS WITH GROUP=2 AND 3 SELECTED BY ROAR
3
4      VARIABLE NO.    VARIABLE NAME
5      1              UP
6      2              4
7      3              52
8      4              XSS
9      5              V3
10     6              SF
11     7              53
12     8              DIG
13     9              F4
14    10             K4R
15    11             VF=10
16    12             VF=4
17    13             VF=20
18    14             VF=20
19    15             NC=10
20    16             NC=4
21    17             SF=20
22    18             D15=JP
23    19             DIG=4
24    20             DIG=4
25    21             D15=SS
26    22             D15=20
27    23             3N=1
28    24             F4=4
29    25             C4=20
30    26             SF=4
31    27             F=2XSS
32
33     DIMENSION PRU(31),IJO(4),IXS(11),STAT(52),IXV(31)
34     DIMENSION INO(31),REST(465,4),NVAR(100)
35     DIMENSION INO(31),4K(10)
36     DIMENSION YN(31),V3(10),XX(100,71),TEMP(31),VC(31,31)
37
38     TAPE1 IS DATA TAPE.
39     DATA IS IN 4F20.10 FORMAT
40     INDEPENDENT VARIABLES ARE FIRST
41     DEPENDENT VARIABLE IS LAST
42     TAPE1 IS CONTROL TAPE
43     NUMBER OF INDEPENDENT VARIABLES
44     NUMBER OF OBSERVATIONS
45     IN 22 FORMAT
46     TX=176
47
48     THIS IS THE MAXIMUM NUMBER OF DATA POINTS, AND
49     SHOULD BE CHANGED IF DIMENSIONS ARE CHANGED.
50
51     POINT 2J=0
52     FORMAT(" MAXIMUM NUMBER OF VARIABLES = 23")
53     P=23 (2,205011VA)
54     FORMAT(11)
55     POINT 2J=1,IVAP
56     FORMAT(" THERE ARE ",I7," INDEPENDENT VARIABLES")

```

Figure 9. Sample Leaps and Bounds



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PAGE 3

11/30/76 18.35.40

FTN 4.6+46

PROGRAM MULTI 74/74 OPI=1

```

115 IF (IER.NE.37) GO TO 5
    PRINT 2012,(I4(I),I=1,KX)
    FORMAT(" VARIABLES DELETED = ",2015)
    ICONT=0
    DO 4, I=1,KX
120 IF (IUC(I).EQ.0) ICONT=ICONT+1
        CONTINUE
        KX=ICONT+1
    C
    C CALL USLEAP(IJ01,47,IK5,STAT,IX4,IV13,IX3,3FST,19)
    IF (IER.NE.0) PRINT 2013,IER
    GO TO 3
125 FORMAT(" ..... ERROR IN USLEAP"// " ERROR NUMBER
        1 " ,IX// " .....")
    END

```

REGRESSIONS WITH 1 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.556119E+02	2
.556439E+02	3

REGRESSIONS WITH 2 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.64515E+02	2 7
.62295E+02	2 3

REGRESSIONS WITH 3 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.63562E+02	2 1 7
.63431E+02	2 1 23

REGRESSIONS WITH 4 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.74902E+02	2 6 7 20
.74005E+02	2 6 7 19

REGRESSIONS WITH 5 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.75431E+02	2 6 7 18 20
.76104E+02	2 6 7 9 20

REGRESSIONS WITH 6 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.72403E+02	2 6 7 3 19 20
.73094E+02	2 6 7 13 20 20

REGRESSIONS WITH 7 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.73404E+02	1 2 6 7 3 19 20
.79012E+02	2 6 7 9 10 20 27



REGRESSIONS WITH 8 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION  
VARIABLES  
1 2 6 7 9 18 21 27  
-017732E+02  
-032922E+02

REGRESSIONS WITH 9 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION  
VARIABLES  
1 2 6 7 9 12 21 26 27  
-021025E+02  
-014034E+02

REGRESSIONS WITH 10 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION  
VARIABLES  
1 2 5 7 9 19 21 23 26 27  
-010316E+02  
-025148E+02

REGRESSIONS WITH 11 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION  
VARIABLES  
1 2 5 6 7 9 19 20 23 26 27  
-013970E+02  
-015727E+02

REGRESSIONS WITH 12 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION  
VARIABLES  
1 2 6 7 13 15 19 20 24 25 26 27  
-049464E+02  
-054432E+02

REGRESSIONS WITH 13 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION  
VARIABLES  
1 2 6 7 13 14 19 20 23 24 25 26 27  
-059330E+02  
-059549E+02

REGRESSIONS WITH 14 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION  
VARIABLES  
1 2 6 7 13 16 19 20 22 23 24 25 26 27  
-044055E+02  
-071203E+02

REGRESSIONS WITH 15 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.85077E+02	1 2 6 7 13 14 15 16 17 18 19 20 22 23 24 25 26 27
.55107E+02	1 2 6 7 13 14 15 16 17 18 19 20 22 23 24 25 26 27

REGRESSIONS WITH 16 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.445761E+02	1 2 6 7 13 14 15 16 17 18 19 20 22 23 24 25 26 27
.316333E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27

REGRESSIONS WITH 17 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.293357E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27
.083311E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27

REGRESSIONS WITH 18 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.133005E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27
.091411E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27

REGRESSIONS WITH 19 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.031727E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27
.031336E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27

REGRESSIONS WITH 20 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.025116E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27
.009025E+02	1 2 6 7 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 21 VARIABLE(S) (ADJUSTED R-SQUARED)

CRITERION	VARIABLES
.017975E+02	1 2 5 6 7 12 13 14 15 16 17 18 19 20 22 23 24 25 26 27
.008333E+02	1 2 5 6 7 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 22 VARIABLE(S) (ADJUSTED R-SQUARED)

[illegible]

REGRESSIONS WITH 23 VARIABLE(S) (ADJUSTED R-SQUARED)

STU:KWA  
404:KWA

REGRESSIONS WITH 24 VARIABLE(S) (ADJUSTED R-SQUARES)

[illegible]

REGRESSIONS WITH 25 VARIABLE(S) (ADJUSTED R-SQUARE)

CRITERION	VARIABLES																					
09: 05V: 02	1	2	3	5	6	7	10	11	12	13	14	15	17	19	20	21	22	23	24	25	26	27
30: 05V: 02	1	2	3	5	6	7	10	11	12	13	14	15	17	19	20	21	22	23	24	25	26	27

REGRESSIONS WITH 26 VARIABLE(S) (ADJUSTED R-SQUARED)

DATE	TIME	WAVELENGTH	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26
1950/10/02	22		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26
1950/10/02	27		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26

REPRESSIONS WITH 27 VARIABLE(S) (ADJUSTED R-SQUARE)

CREATION: 05SEP62  
 VARIAB: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25  
 26 27



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BEST REGRESSIONS WITH 20 VARIABLES (ADJUSTED R-SQUARE)

VARIABLE	COEFFICIENT	PARTIAL F	A <sub>2</sub> MA
1	.713712+10	.11433F+02	.550322-03
2	.300000+00	.852122+01	.122355-01
3	-.200000+01	.113170+02	.503341E-07
4	-.200000+01	.291700+02	.171845-06
5	.401171+01	.590925+01	.223312-01
6	-.200000+00	.210000+01	.154225-00
7	-.200000+00	.153317+02	.261250E-03
8	-.200000+00	.851800+02	.331745-03
9	.200000+00	.312240+01	.503300-01
10	-.200000+00	.332340+01	.373330-01
11	.200000+00	.429300+01	.160133-01
12	-.200000+00	.117080+02	.130733-02
13	-.200000+00	.221700+01	.142135-00
14	.200000+00	.341500+01	.315333-01
15	.200000+00	.250000+01	.111133-00
16	.200000+00	.337200+01	.290333-02
17	.200000+00	.131430+02	.583333-03
18	.200000+00	.213150+02	.100133-00
19	.200000+00	.270330+02	.550300-05
20	-.766325+00	.311290+02	.220300-05

REGRESSIONS WITH 1 VARIABLE(S) (HALLOWS C2)  
CRITERION  
+1.121E+03  
+259.75E+03  
VARIABLES  
2  
3

REGRESSIONS WITH 2 VARIABLE(S) (HALLOWS C2)  
CRITERION  
+129.05E+03  
+134.25E+03  
VARIABLES  
2 7  
2 7

REGRESSIONS WITH 3 VARIABLE(S) (HALLOWS C2)  
CRITERION  
+102.03E+03  
+103.17E+03  
VARIABLES  
2 6 7  
2 7 23

REGRESSIONS WITH 4 VARIABLE(S) (HALLOWS C2)  
CRITERION  
+74.73E+02  
+75.04E+02  
VARIABLES  
2 6 7 20  
2 6 7 19

REGRESSIONS WITH 5 VARIABLE(S) (HALLOWS C2)  
CRITERION  
+67.77E+02  
+53.21E+02  
VARIABLES  
2 6 7 13 20  
2 6 7 9 20

REGRESSIONS WITH 6 VARIABLE(S) (HALLOWS C2)  
CRITERION  
+57.137E+02  
+55.37E+02  
VARIABLES  
2 6 7 9 13 20  
2 6 7 13 20 24

REGRESSIONS WITH 7 VARIABLE(S) (HALLOWS C2)  
CRITERION  
+52.271E+02  
+50.55E+02  
VARIABLES  
1 2 6 7 9 13 20  
2 6 7 9 13 20 24

REGRESSIONS WITH 8 VARIABLE(S) (MALLONS C-)

CRITERION  
VARIABLES  
1 2 6 7 3 14 21 27  
+56822E+02  
+37289E+02

REGRESSIONS WITH 9 VARIABLE(S) (MALLONS C-)

CRITERION  
VARIABLES  
1 2 6 7 3 14 21 26 27  
+39284E+02  
+34114E+02

REGRESSIONS WITH 10 VARIABLE(S) (MALLONS C-)

CRITERION  
VARIABLES  
1 2 6 7 3 14 21 26 27  
+35331E+02  
+37147E+02

REGRESSIONS WITH 11 VARIABLE(S) (MALLONS C-)

CRITERION  
VARIABLES  
1 2 6 7 3 14 21 26 27  
+13256E+02  
+38256E+02

REGRESSIONS WITH 12 VARIABLE(S) (MALLONS C-)

CRITERION  
VARIABLES  
1 2 6 7 3 14 21 23 24 25 26 27  
+21307E+02  
+24101E+02

REGRESSIONS WITH 13 VARIABLE(S) (MALLONS C-)

CRITERION  
VARIABLES  
1 2 6 7 3 14 21 23 24 25 26 27  
+27166E+02  
+23733E+02

REGRESSIONS WITH 14 VARIABLE(S) (MALLONS C-)

CRITERION  
VARIABLES  
1 2 6 7 3 14 21 23 24 25 26 27  
+15305E+02  
+17371E+02



REGRESSIONS WITH 15 VARIABLE(S) (MALLONS CP)

CRITERION  
 .153185E+02  
 .153285E+02  
 1 2 6 7 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27  
 1 2 6 7 12 13 14 15 20 22 23 24 25 26 27

REGRESSIONS WITH 16 VARIABLE(S) (MALLONS CP)

CRITERION  
 .172511E+02  
 .172611E+02  
 1 2 6 7 13 14 17 18 19 20 21 22 23 24 25 26 27  
 1 2 6 7 13 14 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 17 VARIABLE(S) (MALLONS CP)

CRITERION  
 .171432E+02  
 .172271E+02  
 1 2 6 7 13 14 17 18 19 20 21 22 23 24 25 26 27  
 1 2 6 7 13 14 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 18 VARIABLE(S) (MALLONS CP)

CRITERION  
 .162422E+02  
 .173075E+02  
 1 2 5 7 13 14 17 18 19 20 21 22 23 24 25 26 27  
 1 2 6 7 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 19 VARIABLE(S) (MALLONS CP)

CRITERION  
 .156232E+02  
 .173717E+02  
 1 2 5 7 13 14 17 18 19 20 21 22 23 24 25 26 27  
 1 2 6 7 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 20 VARIABLE(S) (MALLONS CP)

CRITERION  
 .157350E+02  
 .173250E+02  
 1 2 5 7 13 14 17 18 19 20 21 22 23 24 25 26 27  
 1 2 6 7 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 21 VARIABLE(S) (MALLONS CP)

CRITERION  
 .173925E+02  
 .174251E+02  
 1 2 5 6 7 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27  
 1 2 5 5 7 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

REGRESSIONS WITH 22 VARIABLE(S) (MALLONS C2)

CRITERION  
VARIABLES  
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27  
-192589E+02  
-203721E+02

REGRESSIONS WITH 21 VARIABLE(S) (MALLONS C2)

CRITERION  
VARIABLES  
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27  
-202241E+02  
-211435E+02

REGRESSIONS WITH 24 VARIABLE(S) (MALLONS C2)

CRITERION  
VARIABLES  
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27  
-225315E+02  
-225420E+02

REGRESSIONS WITH 25 VARIABLE(S) (MALLONS C2)

CRITERION  
VARIABLES  
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27  
-242429E+02  
-247558E+02

REGRESSIONS WITH 26 VARIABLE(S) (MALLONS C2)

CRITERION  
VARIABLES  
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26  
-251135E+02  
-252001E+02  
27

REGRESSIONS WITH 27 VARIABLE(S) (MALLONS C2)

CRITERION  
VARIABLES  
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25  
-263005E+02  
26 27

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BEST REGRESSIONS WITH 16 VARIABLE(S) (MIL-045 CP)

VARIABLE	COEFFICIENT	PARTIAL F	
1	.34410E+00	.29130E+02	1.34A
2	.33261E+00	.73157E+01	.49313E-04
6	-.32670E+01	.13031E+02	.35435E-02
7	-.21251E+01	.83111E+02	.62737E-04
8	-.51930E+01	.63777E+01	.29171E-08
13	-.62247E+00	.15736E+02	.66333E-02
14	.42226E+00	.12714E+02	.79121E-04
17	.78087E+00	.13740E+01	.33434E-03
18	-.15155E+02	.15740E+01	.43314E-02
19	-.16774E+01	.15740E+01	.49115E-02
21	.37290E+01	.37247E+01	.21734E-02
23	.16451E+00	.72751E+02	.13811E-04
24	.21730E+00	.13414E+01	.33753E-02
25	-.51702E+00	.12240E+02	.10435E-02
25	.71150E+00	.71103E+02	.34172E-04
25	.71150E+00	.71103E+02	.10521E-04
27	-.51121E+00	.84947E+02	.39133E-06

..... LJP00FL //// END OF LIST ////



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LEAST SQUARES FIT FOR DATA IN COLUMN 22									
AS A LINEAR FUNCTION OF 24 PREDICTOR VARIABLES IN COLUMNS 1, 2, 3, 4, 5, 6, 7, 8,									
9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21									
USING 15 NON-ZERO WEIGHTS AND 5 ZERO WEIGHTS IN COLUMN 23									
ROW	COL. 1	COL. 2	COL. 3	COL. 4	COL. 5	COL. 6	COL. 7	COL. 8	COL. 9
1	3.32	2.34	0.	0.	0.	0.	0.	0.	0.
2	10.32	3.60	0.	0.	0.	0.	0.	0.	0.
3	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
4	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
5	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
6	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
7	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
8	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
9	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
10	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
11	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
12	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
13	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
14	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
15	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
16	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
17	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
18	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
19	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
20	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
21	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
22	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
23	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
24	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
25	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
26	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
27	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
28	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
29	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
30	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
31	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
32	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
33	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
34	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
35	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
36	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
37	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
38	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
39	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
40	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
41	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
42	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
43	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
44	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
45	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
46	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
47	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
48	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
49	1.33	3.60	0.	0.	0.	0.	0.	0.	0.
50	1.33	3.60	0.	0.	0.	0.	0.	0.	0.

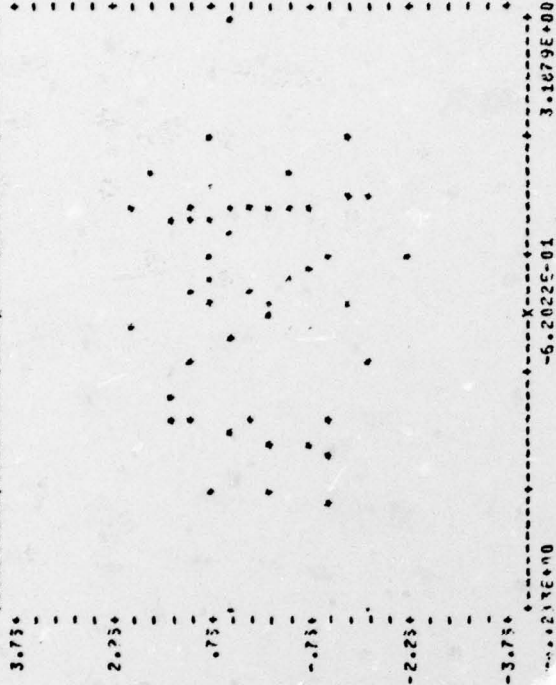
Figure 10. Sample OMNITAB Output

[illegible]

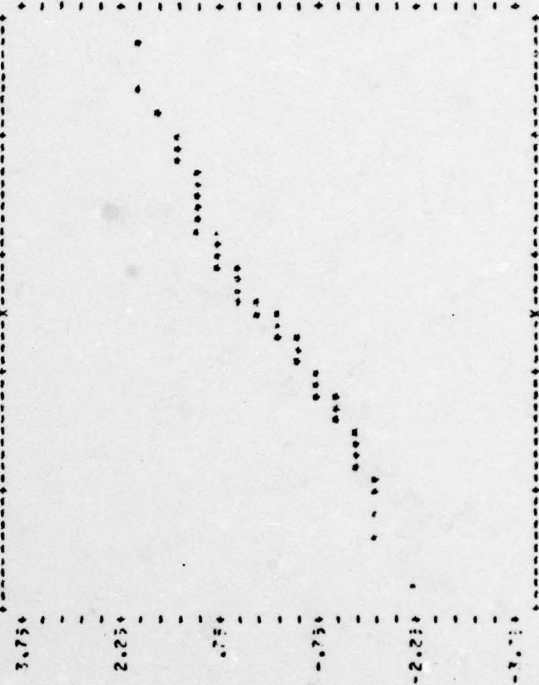
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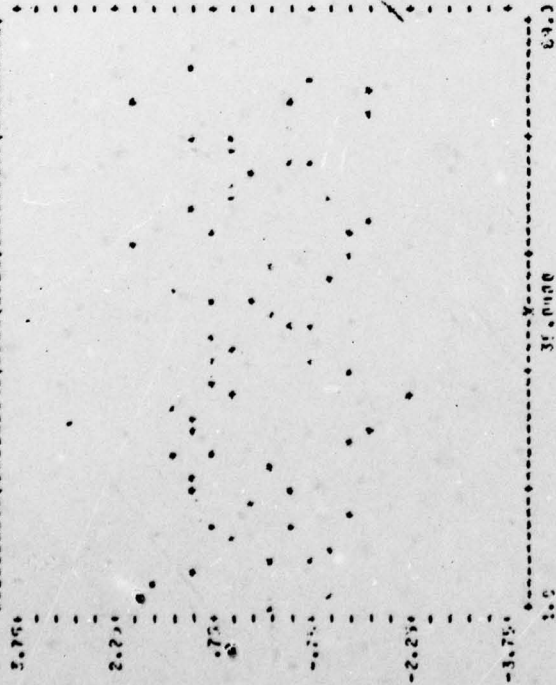
STANDARDIZED RESIDUALS VS PREDICTED VALUES



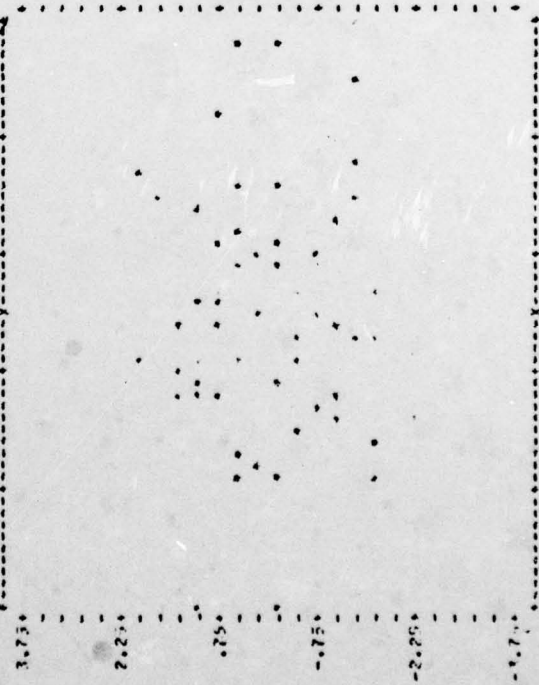
PROBABILITY PLOT OF STANDARDIZED RESIDUALS



STANDARDIZED RESIDUALS VS RUN NUMBER



STANDARDIZED RESIDUALS VS VARIABLE X





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LEAST SQUARES FIT FOR DATA IN COLUMN 22  
AS A LINEAR FUNCTION OF 21 PREDICTOR VARIABLES IN COLUMNS 1, 2, 3, 4, 5, 6, 7, 8,  
9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21  
USING 13 NON-ZERO WEIGHTS AND 6 ZERO WEIGHTS IN COLUMN 23

VARIANCE-COVARIANCE MATRIX OF THE ESTIMATED COEFFICIENTS

CO-UM	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23
1	2523062																						
2	317728419	10758767																					
3	317728419	10758767	317728419																				
4	317728419	10758767	317728419	317728419																			
5	317728419	10758767	317728419	317728419	317728419																		
6	317728419	10758767	317728419	317728419	317728419	317728419																	
7	317728419	10758767	317728419	317728419	317728419	317728419	317728419																
8	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419															
9	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419														
10	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419													
11	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419												
12	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419											
13	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419										
14	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419									
15	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419								
16	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419							
17	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419						
18	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419					
19	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419				
20	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419			
21	317728419	10758767	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419	317728419		

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20 --040969599 .01133027 .001306125 .317773443 -.009222593 .023330642  
21 .012223291 -.011067954 -.002507892 -.011114143 .0074795832 -.017015033 .313074732

ANALYSIS OF VARIANCE  
-DEFINIENT ON ORDER VARIABLES ARE ENTERED, UNLESS VECTORS ARE ORTHOGONAL-

COL-UM	SS=ED. DUE TO CCE.	CUM. MS REJECTION	D.F.	CUM. RESIDUAL MS	D.F.	F(COEF=0)	P(F)	F(COEF=0)	P(F)
1	0.022046	0.022046	1	2.327117	62	35.107	.000	29.249	.000
2	26.127705	26.127705	2	1.312722	61	143.761	.000	28.042	.000
3	4.222028	30.349733	3	1.131652	60	210.154	.000	22.923	.000
4	3.874594	34.224327	4	.991424	59	31.159	.000	12.233	.000
5	15.137501	49.361828	5	.812103	58	60.236	.000	10.771	.000
6	4.121056	53.482884	6	.693377	57	10.523	.000	7.679	.000
7	2.021206	55.504090	7	.580651	56	9.173	.007	7.092	.000
8	2.078711	57.582801	8	.481642	55	10.059	.003	7.014	.000
9	1.100022	58.682823	9	.386231	54	1.510	.839	6.773	.000
10	1.074077	60.756899	10	.341056	53	1.284	.145	6.935	.000
11	.550174	62.307073	11	.288831	52	1.451	.230	7.210	.000
12	.320413	63.627486	12	.240213	51	1.547	.253	7.780	.000
13	.01105445	63.638540	13	.237115	50	1.007	.940	8.502	.000
14	1.078740	64.717280	14	.210754	49	1.004	.609	9.504	.000
15	1.164500	65.881780	15	.206916	48	10.593	.000	6.573	.000
16	.01205077	65.893830	16	.202231	47	.008	.824	6.775	.000
17	.7070502	66.600880	17	.197972	46	3.000	.090	10.520	.000
18	1.003027	67.603907	18	.192750	45	7.183	.010	12.393	.000
19	.841057	68.444964	19	.186127	44	1.151	.210	16.139	.000
20	.278029	68.722993	20	.178512	43	2.573	.119	20.413	.000
21	3.614153	72.337146	21	.169908	42	34.153	.000	19.163	.000
RESIDUAL	16.594791		42						
TOTAL	105.37418		103						

COMPUTER

LEAST SQUARES FIT FOR DATA IN COLUMN 22  
 AS A LINEAR FUNCTION OF 21 EXPLANATORY VARIABLES IN COLUMNS 1, 2, 3, 4, 5, 6, 7, 8,  
 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21  
 USING 10 NON-WEIGHTED OBSERVATIONS AND 10 ZERO WEIGHTS IN COLUMN 23

ESTIMATES FROM LEAST SQUARES FIT				FIT OMITTING LAST COLUMN			
COLUMN	COEFFICIENT	S.D. OF COEFF.	RATIO	COEFFICIENT	S.D. OF COEFF.	RATIO	
1	-.0315175	.001772	-0.04	-4.5022532	.0022757	-6.59	
2	.0170079	.0013203	3.01	.2561218	.1113517	2.29	
3	.0010029	.0005021	2.02	.7710317	.1003201	1.06	
4	.0000000	.0000000	0.00	-2.1205315	1.0255312	-2.08	
5	.0000000	.0000000	0.00	-2.0716313	.0721850	-5.45	
6	.0000000	.0000000	0.00	3.1157736	2.0283917	1.16	
7	.0000000	.0000000	0.00	.037005559	.1105509	.18	
8	.0000000	.0000000	0.00	-.1359189	.1613564	-2.66	
9	.0000000	.0000000	0.00	.3055404	.2295176	1.32	
10	.0000000	.0000000	0.00	.1243323	.1763519	.70	
11	.0000000	.0000000	0.00	-.2307071	.2007128	-1.15	
12	.0000000	.0000000	0.00	.1757230	.1578917	1.12	
13	.0000000	.0000000	0.00	-.7117036	.1751742	-1.50	
14	.0000000	.0000000	0.00	-1.1832076	.0641552	-1.25	
15	.0000000	.0000000	0.00	3.7192619	1.0259359	2.29	
16	.0000000	.0000000	0.00	-.05193217	.0519320	-.05	
17	.0000000	.0000000	0.00	.19400195	.0907233	1.05	
18	.0000000	.0000000	0.00	.08917215	.2245154	.40	
19	.0000000	.0000000	0.00	-.14150402	.1375254	-.03	
20	.0000000	.0000000	0.00	-.091994336	.076327322	-1.26	
21	.0000000	.0000000	0.00				

RESIDUAL STANDARD DEVIATION = .00192533  
 BASED ON 10 OBSERVATIONS OF 21 VARIABLES  
 THE NUMBER OF CORRECTLY COMPUTED DIGITS IN EACH COEFFICIENT USUALLY DIFFERS BY LESS THAN 1 FROM THE NUMBER GIVEN HERE  
 \$ ADJUST FIT

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LAST OF COMP. RES. DATA AND DIAGNOSTICS

124

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### Vita

Larry James Pulcher was born on September 18, 1951 in Belleville, Illinois. He graduated from high school in Dupon, Illinois, whereupon he attended Parks College of Aeronautical Technology of St. Louis University. He earned a degree there in Aerospace Engineering in 1972 and was commissioned into the Air Force through the ROTC program. His first assignment was as a Minuteman Launch Control Officer at Minot AFB, North Dakota where he was awarded a Regular commission and the Air Force Commendation Medal.

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		6. PERFORMING ORG. REPORT NUMBER
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19. KEY WORDS (Continue on reverse side if necessary and identify by block number)  Multiple Regression  Selection of Variables		
20. ABSTRACT (Continue on reverse side if necessary and identify by block number)  The Test of Equality between Subsets of Coefficients in Two Regressions is developed and applied as a means to pre-screen variables from a regression model.  Some criterion for selection of variables are discussed and some existing regression packages are applied to data on characteristics of avionics equipment for comparison purposes.		

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